

Decrease in climatic conditions favouring floods in the south-east of Belgium over 1959–2010 using the regional climate model MAR

Coraline Wyard,^{a*} Chloé Scholzen,^a Xavier Fettweis,^a Jean Van Campenhout^b and Louis François^c

^a *Laboratory of Climatology and Topoclimatology, University of Liège, Belgium*

^b *Laboratoire d'Hydrographie et de Géomorphologie Fluviale, University of Liège, Belgium*

^c *Unité de Modélisation du Climat et des Cycles Biogéochimiques, University of Liège, Belgium*

ABSTRACT: The Ourthe River, in the south-east of Belgium, has a catchment area of 3500 km² and is one of the main tributaries of the Meuse River. In the Ourthe, most of the flood events (FE) occur during winter and about 50% of them are due to heavy rainfall events combined with an abrupt melting of the snowpack covering the Ardennes massif during winter. This study aims to determine whether trends in extreme hydroclimatic events generating floods can be detected over the last century in Belgium, where a global warming signal can be observed. Hydroclimatic conditions favourable to floods were reconstructed over 1959–2010 using the regional climate model MAR ('Modèle Atmosphérique Régional') forced by the ERA-Interim/ERA-40, the ERA-20C and the NCEP/NCAR-v1 reanalyses. Extreme run-off events, which could potentially generate floods, were detected using run-off caused by precipitation events and snowpack melting from the MAR model. In the validation process, the MAR-driven temperature, precipitation and snow depth were successfully compared to daily weather data over the period 2008–2014 for 20 stations in Belgium. MAR also showed its ability to detect up to 90% of the hydroclimatic conditions which effectively generated observed floods in the Ourthe River over the period 1974–2010. Conditions favourable to floods in the Ourthe River catchment present a negative trend over the period 1959–2010 as a result of a decrease in snow accumulation and a shortening of the snow season. This trend is expected to accelerate in a warmer climate. However, regarding the impact of the extreme precipitation events evolution on conditions favouring floods, the signal is less clear because the trends depend on the reanalysis used to force the MAR model.

KEY WORDS regional climate models; Europe; Belgium; Ourthe River; floods; snow; precipitation

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1. Introduction

Changes in seasonal snow cover cause concern in many mid- to high-latitudes countries because of its implications in water supply and flood risks (e.g. Barnett *et al.*, 2005; Beniston, 2012a). Over the last decades, a significant general trend toward decreasing snow depth, snow duration and snowfall amount with interdecadal variability has been identified in various European regions such as the Alps (e.g. Durand *et al.*, 2009; Valt and Cianfarra, 2010; Beniston, 2012b), Britain (e.g. Kay, 2016), Norway (e.g. Skaugen *et al.*, 2012; Dyrødal *et al.*, 2013), or Eastern Europe (e.g. Falarz, 2004; Brown and Petkova, 2007; Birsan and Dumitrescu, 2014). This declining snow cover is responsible for a decrease in the intensity and in the frequency of the floods dominated by snowmelt which is predicted to accelerate in the future (e.g. Bell *et al.*, 2016; Vormoor *et al.*, 2016).

In Belgium, a low-lying country of western Europe [maximum elevation = 694 m above sea level (a.s.l.)], snow can cover the summits of the Ardennes massif from 1 to 2 months per year on average, and can reach 80 cm with consequences on water management, biodiversity and tourism activities. When combined with heavy rainfall events, the rapid melting of this snow cover is responsible for major floods in the Meuse River catchment, located in the south-east of Belgium. However, despite the aforementioned implications, very few studies about the recent evolution of seasonal snow cover have been carried out in Belgium. Moreover, these studies, which do not cover the last two decades (1990–2015), have been conducted either over much shorter periods (Sneyers, 1967a, 1967b; Erpicum *et al.*, 1991) or on a more limited number of weather stations (Sneyers, 1965) to build robust statistics.

Therefore, this research aims to assess whether seasonal snow cover and climatic conditions favouring floods have changed over the last 50 years in Belgium. For this purpose, the regional climate model (RCM) MAR (for 'Modèle Atmosphérique Régional' in French) was used over the period 1959–2010, forced by three reanalyses. As snow

* Correspondence to: C. Wyard, Laboratory of Climatology and Topoclimatology, University of Liège, Quartier Village 4, Clos Mercator 3, 4000 Liège, Belgium. E-mail: Coraline.Wyard@ulg.ac.be

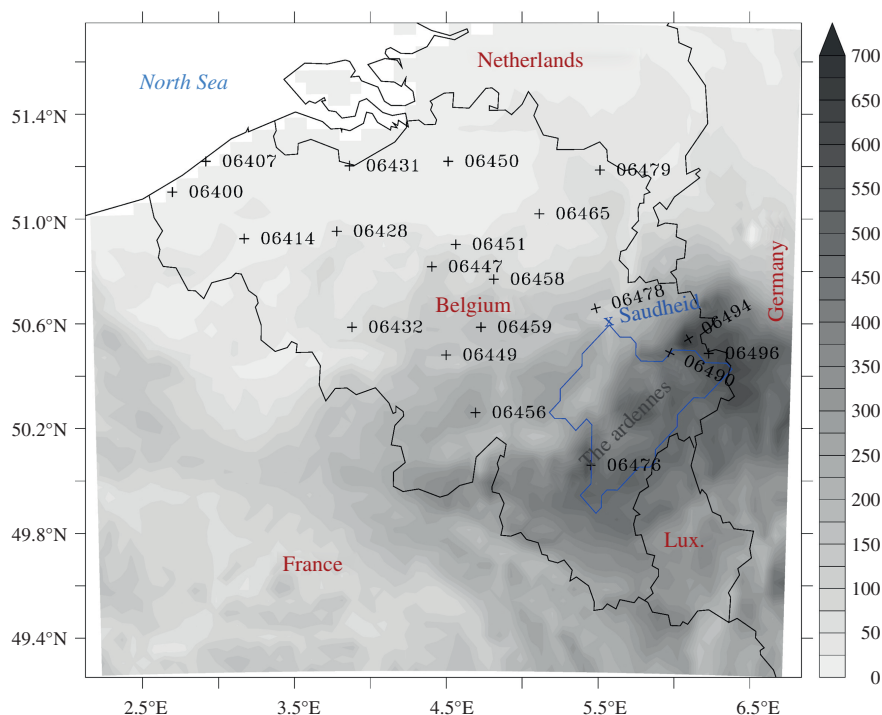


Figure 1. Orography of the study area, location of the weather stations used in this study with their WMO identifier, limits of the Ourthe River catchment area upstream of the gauging station of Sauheid. [Colour figure can be viewed at wileyonlinelibrary.com.]

cover records over the Belgian territory are patchy, discontinuous or made with techniques that change over time, a RCM like MAR, which was specially designed to model snow (Gallée *et al.*, 2001; Fettweis *et al.*, 2013), is a highly valuable tool for studying the recent changes in snow cover at a high-spatial resolution. The other RCM running over Belgium, such as the ALADIN and AROME combined model (ALARO-0 model), are used to model extreme events in summer such as Urban Heat Island (Hamdi *et al.*, 2014) and precipitation (De Troch *et al.*, 2013) while we are focussing here on the winter season and snow.

After a brief description of the study area, the RCM used and the validation data are described in Section 2. Section 3 first presents and discusses the results of the evaluation of the MAR model over the Belgian territory by comparing the model outputs with daily observations from 20 weather stations of the surface synoptic observations (SYNOP) network over the period 2008–2014. Secondly, Section 3 details the statistical analysis performed to study the evolution of (i) the seasonal snow cover simulated by MAR over the period 1959–2010 and, (ii) the conditions favourable to floods in the catchment area (3500 km²) of the Ourthe River in the Ardennes region/massif. Conclusions and prospects are finally reported in Section 4.

2. Data and methods

2.1. Study area

This study focuses on the Belgian territory (see Figure 1), and more specifically on its south-eastern part where the presence of the Ardennes massif implies a hilly relief. The

summits of the Ardennes consist in high plateaus covered with spruce forests and heathland. The highest summit is the Hautes Fagnes plateau whose altitude reaches 694 m a.s.l. Among the several rivers which drain the Ardennes massif, the largest one is the Ourthe River, with a catchment area of about 3500 km² (Pauquet and Petit, 1993). This river is also the main tributary of the Meuse River, which is one of the largest rivers of Western Europe, with a catchment area of about 36 000 km² and a length of 950 km. While the Ourthe River is mainly rain-fed, the snowpack that covers the Ardennes summits during winter also influences its discharge because of its buffering effect (Driessen *et al.*, 2010). The Ourthe reaches its highest level either in winter or in spring when the snowpack is melting. About half of the observed floods are due to abundant rainfall alone. However, the rapid melting of the snowpack that covers the Ardennes summits, combined with heavy rainfall events, are responsible for major floods in the lower part of the Ourthe River, as it was the case in February 1984, March 1988, December 1991 or, more recently, in January 2002 and January 2011 (Pauquet and Petit, 1993; de Wit *et al.*, 2007).

Westerlies bring precipitation in the Ardennes throughout the year, and precipitation is all the more abundant as altitude is high. For example, while the mean annual precipitation amount is on average around 800 mm year⁻¹ in the Low and Middle-Belgium, it can reach 1400 mm year⁻¹ on the Hautes Fagnes plateau. In general, the climate in the high plateaus of the Ardennes is wetter, colder and more snowy than in the rest of Belgium. Snowfall events can generate a snow cover that can last from 1 to 2 months per year on average,

and that can reach up to 70–80 cm for the most snowy winters.

This peculiar climate which characterizes the summits of Belgium, especially in the Hautes Fagnes plateau, allows the existence of a remarkable ecosystem at this range of altitude (between 500 and 700 m a.s.l.) and latitude (between 49° and 51°N) (Plunus *et al.*, 2014). This ecosystem is characterized by active bogs, wet heathland, alpine and even subarctic animal and plant species. Most of these plant species can survive to extreme cold temperatures in the winter months thanks to the protective snow cover that acts as a thermal insulator. Both the peculiar climate and ecosystem of the Belgian high plateaus also generate tourism activities especially in winter (skiing, hiking, educative walks).

2.2. The MAR model description and experiments

The RCM used in this study is the MAR model which, although being initially designed for polar regions (Gallée and Schayes, 1994; Fettweis *et al.*, 2013), was also applied to temperate (Brasseur, 2001; Brasseur *et al.*, 2002) and tropical regions (Brasseur *et al.*, 1998; De Ridder and Gallée, 1998; Gallée *et al.*, 2004). MAR is a hydrostatic primitive equation model in which convection is parametrized according to Bechtold *et al.* (2001). The atmospheric part of MAR is completely described in Gallée and Schayes (1994) and Gallée (1995). Furthermore, MAR is coupled to the 1-D surface vegetation atmosphere transfer scheme SISVAT (Soil Ice Snow Vegetation Atmosphere Transfer) which is detailed in De Ridder and Gallée (1998). The snow-ice part of SISVAT is the snow model CROCUS from the CEN (Centre d'Etudes de la Neige) described in Brun *et al.* (1992). The MAR-SISVAT coupling allows the consideration of three sub-pixel surface characteristics for a same MAR pixel. The coupling also allows interaction between surface and atmosphere (energy and moisture transfers), snow accumulation and snow melting on the surface, water percolation into the soil/snow, and run-off of exceeding water.

In this study, the version 3.6 of MAR, which is currently being developed at the Laboratory of Climatology and Topoclimatology of the University of Liège (Liège, Belgium), was adapted to the Belgian climate. Simulations were performed at a resolution of 5 km over a domain of 120×110 pixels centred on (4.3°W; 50.4°N) (see Figure S1, Supporting information). Boundary conditions [temperature, wind, humidity and pressure at each pressure level of MAR, and sea surface temperature (SST)] were provided every 6 h by forcing reanalyses. We used the third generation ERA-Interim reanalysis ($0.75^\circ \times 0.75^\circ$ horizontal resolution and 60 vertical levels from the surface to 0.01 hPa) (Dee *et al.*, 2011) from the European Centre for Medium-Range Weather Forecasts (ECMWF) over 1979–2014, completed by the second generation ERA40 reanalysis ($1.125^\circ \times 1.125^\circ$ horizontal resolution and 60 vertical levels from the surface to 0.01 hPa) (Uppala *et al.*, 2005) over 1958–1978. The first generation NCEP/NCAR-v1 reanalysis ($2.5^\circ \times 2.5^\circ$

horizontal resolution and 28 vertical levels from the surface to 3 hPa) from the National Centres for Environmental Prediction/National Centre for Atmospheric Research (NCEP/NCAR) was also used (Kalnay *et al.*, 1996) to force MAR over 1958–2014. In addition, MAR was also forced by the third generation ERA-20C reanalysis ($1.125^\circ \times 1.125^\circ$ horizontal resolution and 91 vertical levels from the surface to 0.01 hPa) from the ECMWF (Poli *et al.*, 2013) over 1958–2010. Land cover was derived from the CORINE (Coordination of the information on the Environment) land cover database from the EEA (European Environment Agency). Regarding the other surface characteristics prescribed in SISVAT, such as vegetation seasonality [provided by the monthly normalized difference vegetation index (NDVI)], vegetation cover fraction [derived from the leaf area index (LAI)] or soil type, we used the same datasets as Gallée *et al.* (2004). Finally, since the first year of our simulations (1958) is used as spin-up, only results after 1959 are considered in our analyses.

2.3. Evaluation datasets

The observations (near-surface temperature, precipitation amount and snow depth) required for the MAR model evaluation were extracted from SYNOP data (<http://www.orgimet.com>) emitted every hour by 20 weather stations spread over the Belgian territory. These weather stations are listed in Table 1 and mapped in Figure 1. As the Mont Rigi station [identified by its World Meteorological Organization (WMO) code 06494 in Table 1 and Figure 1] is the reference station for the Hautes Fagnes plateau, we used snow depth measured at 0000 UTC by a snow depth laser sensor at this weather station. Contrary to manual snow depth measurements, laser sensor-driven snow depth measurements are continuously performed with the same instruments, at the same exact place and, in the case of the Mont Rigi station, within a flat, treeless open plateau, thereby limiting potential site effects which could affect those measurements. However, as SYNOP data from the Mont Rigi station were only available from the end of 2007 onwards, the evaluation period only extends from December 2007 to November 2014 for the 20 weather stations considered in this study.

In order to identify FE in the lower part of the Ourthe River, daily flow rates measured in the river flow gauging station of Sauheid (50.6°N; 5.6°W) (see Figure 1) were extracted for the period 1974–2010 from the database of the 'Direction Générale Opérationnelle de la Mobilité et des Voies Hydrauliques de la Région Wallonne'. The Sauheid station was chosen because of its most downstream location in the Ourthe River before the confluence with the Vesdre and the Meuse. According to Pauquet and Petit (1993), a flow rate of $300 \text{ m}^3 \text{ s}^{-1}$ is considered as the flow rate above which the Ourthe River can overflow.

For all the datasets used during the validation process, gaps in the observations were not filled, and missing data were omitted in our statistics computations.

Table 1. Weather stations from the SYNOP network used in the validation process, with their WMO code, their true location (Lat. 1, Lon. 1 and Alt. 1) and the location of the centre of the nearest MAR pixel (Lat. 2, Lon. 2, Alt. 2).

WMO code	Name	Lat. 1 (°)	Lon. 1 (°)	Alt. 1 (m)	Lat. 2 (°)	Lon. 2 (°)	Alt. 2 (m)
06400	Koksijde	51.08	2.65	4	51.11	2.65	1
06407	Oostende Airport	51.20	2.87	4	51.20	2.86	0
06414	Beitem	50.91	3.12	25	50.89	3.09	24
06428	Munte	50.93	3.73	55	50.94	3.73	37
06431	Gent/Industrie-Zone	51.18	3.82	10	51.21	3.80	5
06432	Chievres	50.57	3.83	60	50.58	3.80	63
06447	Uccle	50.80	4.36	101	50.80	4.37	101
06449	Charleroi/Gosselies	50.46	4.45	187	50.49	4.44	159
06450	Antwerpen/Deurne	51.20	4.47	12	51.21	4.44	12
06451	Bruxelles National	50.88	4.52	55	50.89	4.51	61
06456	Florennes	50.24	4.65	279	50.26	4.65	286
06458	Beauvechain	50.75	4.77	105	50.76	4.80	120
06459	Ernage	50.57	4.68	155	50.58	4.65	158
06465	Schaffen	51.00	5.07	30	50.98	5.09	33
06476	Saint-Hubert	50.04	5.41	563	50.03	5.42	563
06478	Bierset	50.63	5.45	186	50.62	5.43	141
06479	Keine-Brogel	51.17	5.47	55	51.16	5.45	61
06490	Spa/La Sauveniere	50.48	5.92	470	50.48	5.93	484
06494	Mont-Rigi	50.51	6.08	671	50.52	6.07	671
06496	Elsenborne	50.47	6.18	564	50.47	6.21	576

3. Results and discussion

3.1. Evaluation of the MAR model

Values computed by MAR forced by ERA-Interim (called MAR-ERA hereafter), by MAR forced by ERA-20C (called MAR-ERA-20C hereafter) and by MAR forced by NCEP/NCAR-v1 (called MAR-NCEP1 hereafter) are compared over 2008–2014 to daily observations from the 20 weather stations listed in Table 1 and showed in Figure 1. It should be noticed that the MAR-ERA-20C evaluation period only extends from 2008 to 2010 because ERA-20C is not available after 2010. For each weather station, the closest pixel to the location of the station is chosen to represent the MAR outputs. This comparison shows the ability of MAR to simulate the daily variability of the Belgian climate, especially in winter and in the highest parts of the country.

Regarding the *daily mean temperature* (T_{ave}), MAR accurately simulates the T_{ave} variability in winter at Mont Rigi (the highest Belgian weather station), with correlation coefficients above 0.95 and root mean square error (RMSE) values lower than 40% of the observed T_{ave} standard deviation (see Figure 2(a)). MAR-ERA provides slightly better results than MAR-NCEP1 and MAR-ERA-20C with better correlation coefficient and RMSE. MAR tends to underestimate T_{ave} in winter with daily biases ranging from +0.01 to -0.61°C at Mont Rigi following the used forcing. The statistics computed for the 20 weather stations confirm these results and show the ability of MAR to accurately simulate the T_{ave} variability whatever the season (see Figure S2(a)–(d)). They also show that MAR tends to overestimate T_{ave} in spring, autumn, and particularly in summer with biases ranging from +0.95 to $+1.84^{\circ}\text{C}$ (see Figure S2(c)). Consequently, MAR overestimates T_{ave} with annual daily mean biases of

$+0.11^{\circ}\text{C}$ (MAR-ERA-20C), $+0.59^{\circ}\text{C}$ (MAR-ERA) and $+0.70^{\circ}\text{C}$ (MAR-NCEP1) (see Table 2).

Regarding the *daily precipitation amount* (PPN) at Mont Rigi, MAR accurately simulates the PPN variability in winter (dominated by stratiform precipitation) with correlation coefficients above 0.70 and RMSE values lower than 100% of the observed PPN standard deviation (see Figure 2(b)). MAR-ERA provides again the best results. However, the model overestimates PPN in winter with daily mean biases ranging from +23.2% and +40.0% at Mont Rigi. Nevertheless, the statistics computed for the 20 weather stations show that MAR fails to adequately simulate the PPN variability during the other seasons (see Figure S2(f)–(h)). Moreover, MAR tends to underestimate (convective) PPN in summer with biases ranging from -0.1 to -12.9% (see Figure S2(g)). Consequently, MAR slightly underestimates PPN with annual daily mean biases valued at -0.8% (MAR-ERA-20C), -4.3% (MAR-ERA) and -6.4% (MAR-NCEP1) (see Table 3).

Comparison with *snow height* (SH) measured at the Mont Rigi station in winter (see Figure 2(c)) indicates that MAR simulates very well the daily SH variability, with correlation coefficients larger than 0.80 and RMSE values below 65% of the observed SH standard deviation. However, MAR underestimate SH with biases of around -6 cm day^{-1} which is large compared with the daily mean snow depth observed at 15 cm. As there is no temperature biases in winter at Mont Rigi (see Figure 2(a)), and because MAR overestimates wintertime daily precipitation amounts at Mont Rigi (see Figure 2(b)), these negative daily mean biases in SH could be due to the too large fresh snowfall density (300 kg m^{-3}) used in MAR. A fresh snowfall density of 300 kg m^{-3} is typical in polar regions, but in temperate climates, lower values ($\sim 100\text{ kg m}^{-3}$) are more relevant (Judson and Doesken, 2000).

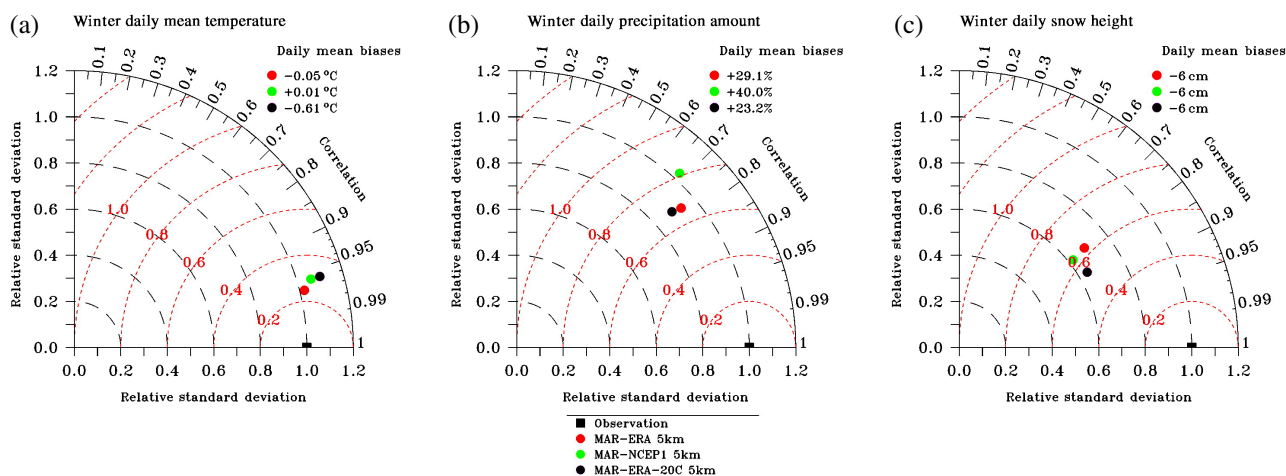


Figure 2. Taylor diagrams showing relative differences between MAR-ERA, MAR-NCEP1 and MAR-ERA-20C in comparison with wintertime observations of (a) daily mean temperature, (b) daily precipitation amount and (c) daily snow height, carried out over 2008–2014 (2008–2010 for MAR-ERA-20C) at the Mont Rigi weather station. [Colour figure can be viewed at wileyonlinelibrary.com].

Table 2. A total of 20 stations averaged statistics of the daily mean temperature observations and model results over the period 2008–2014: correlation (R), standard deviation (σ), root mean squared error (RMSE), daily mean biases and annual daily mean temperature.

	R	σ (°C)	RMSE (°C)	Mean biases (°C)	Mean (°C)
OBS	–	6.40 (6.73)	–	–	10.30 (9.69)
MAR-ERA	0.97	7.22	1.47	+0.59	10.89
MAR-NCEP1	0.96	7.23	1.65	+0.70	11.00
MAR-ERA-20C	(0.96)	(7.51)	(1.42)	(+0.11)	(9.80)
ERA-Interim	0.99	6.42	0.93	+0.32	10.62
NCEP/NCAR-v1	0.95	6.65	1.71	+0.70	11.00
ERA-20C	(0.95)	(6.50)	(1.55)	(–0.24)	(9.45)
Percentage of usable observations: 98.9 %					

The values under brackets indicate that the evaluation period only extends from 2008 to 2010.

Finally, the ability of the MAR model to simulate *climatic conditions favourable to floods* in the Ourthe catchment is assessed by comparing observed flood days with potential flood days and periods computed from the MAR results. On the one hand, 28 FE were identified between January 1974 and December 2010 on the basis of flow rates measured in Sauheid (flooding flow rate $> 300 \text{ m}^3 \text{ s}^{-1}$), which represents a total of 67 flood days. On the other hand, in order to identify potential flood periods, run-off (RU) computed by MAR is integrated over the Ourthe catchment upstream of Sauheid, which represents an area of $\sim 2900 \text{ km}^2$ (see Figure 1). RU is then averaged over the 2 days preceding the flood event (RU2), because the typical response time of the Ourthe River to precipitation ranges from 1 to 2 days (Hazenbergh *et al.*, 2008). The MAR-driven run-off partly consists of precipitation water and of water from melting snow that flow on the ground because of the saturation of the soil in the MAR model. In order to only consider the part of RU2 which really contributes to the river discharge, evaporation computed by MAR is removed from RU2 (RU2_e). At last, as there are two types of floods in the Ourthe River, we also distinguish two types of RU2_e: RU2_e produced by precipitation alone (see the red/dark grey dots in Figure 3)

and RU2_e produced by precipitation combined with snow-pack melting (see the blue/light grey dots in Figure 3). We consider 1 day as favourable to floods if RU2_e is statistically extreme, i.e. if RU2_e is larger than its 95th percentile computed over 1961–1990. The 95th percentile is used because it is commonly used in the literature to define extreme event. Moreover, the trends computed in Section 3.2.2. do not change following the considered percentile (P90, P95, P99 or P99.5). Thereby, about 90% (60/67) of the observed flood days correspond to potential flood days computed by MAR-ERA and MAR-ERA-20C (see Figure 3(a) and (c)), while this proportion only reaches $\sim 60\%$ (41/67) with MAR-NCEP1 (see Figure 3(b)). Similarly, results in terms of FE indicate that about 90% (25/28) of the observed floods correspond to periods identified as favourable to floods according to MAR-ERA-20C and MAR-ERA. By contrast, this proportion only reaches 64% (18/28) with MAR-NCEP1. However, MAR largely overestimates the number of days which could have generated floods over 1974–2010 with excess of 327, 434 and 702 days for MAR-NCEP1, MAR-ERA and MAR-ERA-20C, respectively (see Figure 3). Indeed, floods also depend on non-climatic factors which are not represented in our estimation, such

Table 3. A total of 20 stations averaged statistics of the daily precipitation amount observations and model results over the period 2008–2014: correlation (R), standard deviation (σ), root mean squared error (RMSE), daily mean biases and annual daily mean precipitation amount.

	R	σ (mm day ⁻¹)	RMSE (mm day ⁻¹)	Mean biases (mm day ⁻¹)	Mean (mm day ⁻¹)
OBS	–	4.85 (5.02)	–	–	2.30 (2.26)
MAR-ERA	0.59	3.47	4.03	–0.10 (– 4.3%)	2.20
MAR-NCEP1	0.49	3.28	4.39	–0.15 (– 6.4%)	2.15
MAR-ERA-20C	(0.55)	(3.82)	(4.35)	(–0.02 (–0.8%))	(2.24)
ERA-Interim	0.73	3.44	3.46	–0.17 (– 7.2%)	2.13
NCEP/NCAR-v1	0.56	2.96	4.17	–0.50 (–21.7%)	1.80
ERA-20C	n.a.	n.a.	n.a.	n.a.	n.a.
Percentage of usable observations: 95.6 %					

n.a., the data were not available to compute the statistic analysis. The values under brackets indicate that the evaluation period only extends from 2008 to 2010.

as the initial state of the soil prior to an intense run-off event, the infiltration of water into the water tables, the evolution of the surface permeability over time (as a consequence of the conversion of natural land surfaces into impervious surfaces, due to urbanization), the surface, subsurface and underground flow. The coupling between the climate model MAR and a hydrological model would allow a better representation of these mechanisms and thus a better comparison with observations. However, the aim of this work is not to study the floods evolution but only the climatic conditions favouring floods.

In view of these results, we can conclude that whatever the reanalysis forcing the MAR model, the climate variability is well represented by the model over the Belgian territory, although MAR-ERA seems more fitting than MAR-ERA-20C and particularly MAR-NCEP1. MAR-ERA-20C and MAR-NCEP1 overall exhibit lower correlations and higher RMSE values than MAR-ERA in comparison with observed daily temperature and precipitation. In addition, MAR-ERA allows to find more observed floods than MAR-ERA-20C and MAR-NCEP1. We explain the better results of MAR-ERA by the difference between the assimilation system of all three reanalyses (Kalnay *et al.*, 1996; Dee *et al.*, 2011; Poli *et al.*, 2013). Indeed, all the available data are assimilated in ERA-Interim. All except satellite data are assimilated in NCEP/NCAR-v1, while only surface pressure and surface marine winds are assimilated in ERA-20C. Moreover, NCEP/NCAR-v1 horizontal and vertical resolution is coarser than the resolution of ERA-20C and particularly ERA-Interim. Furthermore, when directly comparing all three reanalyses with daily mean temperature and daily precipitation amount from weather stations (see Figure S2), the correlations and the RMSE values are better than those obtained with our MAR model forced by the same reanalyses. Actually, the weather station-based data used in this study is the same data that had been assimilated into the reanalyses (except ERA-20C), which explains the dependence between the reanalyses and the data. Nevertheless, because of the low-spatial resolution of the reanalyses, MAR forced by the reanalyses exhibits lower biases especially in winter (see Figures S2(a) and (e)). It is therefore better to use the MAR model instead of the raw

reanalyses, which in addition do not adequately simulate the snow pack.

Besides the biases due to the forcing conditions, some discrepancies can also be related to the MAR model itself. For instance, as a result of the underestimation of the cloud cover in MAR, MAR underestimates temperature in winter and overestimate it in summer. Moreover, daily PPN variability is better reproduced in winter (when stratiform precipitation dominates) than in summer (when convective precipitation dominates). Surface and soil properties could also contribute to the MAR model biases through their reflectivity or their water exchanges with the atmosphere.

3.2. Trend analysis over the period 1959–2010

3.2.1. Changes in the Belgian climate

A linear trend analysis is performed for the period 1959–2010 over the entire Belgian territory. This period is the longest common period between all three reanalyses forcing the MAR model. Trend significance is assessed by using the uncertainty range of Snedecor for the 95th confidence interval (Snedecor and Cochran, 1967) (Appendix S1) as used by Fettweis (2007). The trend of a time series is significant if its value is higher than its uncertainty range, which mainly corresponds to the interannual variability of the considered time series. The trend analysis particularly focuses on the winter period [December–January–February (DJF)], as more than 70% of the floods in the Ourthe River occur during this season. Only the trends computed using the MAR-ERA results are displayed because the evaluation of the MAR model showed that the MAR-ERA results are the closest to observations (see Section 3.1.). The trends computed using the MAR-ERA-20C results and the MAR-NCEP1 results are however available from Figure S5 to Figure S14.

Mean winter temperature exhibits an increasing trend ranging from $+0.6^{\circ}\text{C} (52 \text{ years})^{-1}$ to $+1.0^{\circ}\text{C} (52 \text{ years})^{-1}$ in the Ourthe catchment (see Figure S3(a)). However, the mean winter temperature trend is non-significant while mean annual temperature exhibits a significant increasing trend between $+0.8^{\circ}\text{C} (52 \text{ years})^{-1}$ and $+1.0^{\circ}\text{C} (52 \text{ years})^{-1}$ in the Ourthe catchment (see Figure S3(b)). Trends in the winter extreme minimum

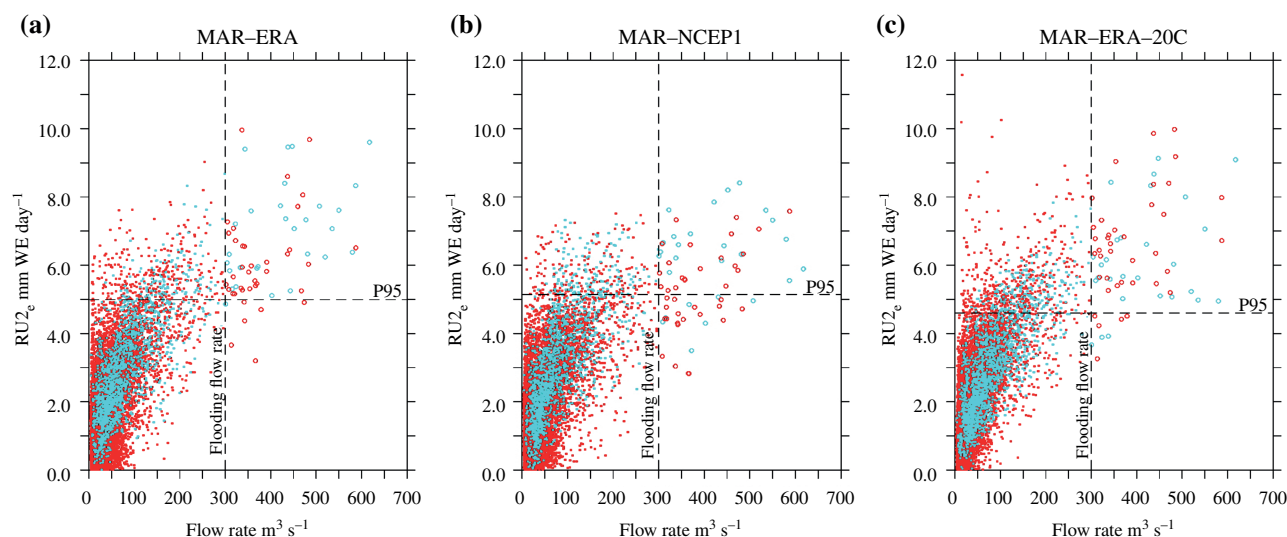


Figure 3. Two-day averaged run-off integrated over the Ourthe River catchment area upstream of Sauheid ($RU2_e$) computed from (a) MAR-ERA, (b) MAR-NCEP1 and (c) MAR-ERA-20C, in relation to the flow rate measured in Sauheid. Each point represents 1 day between 1 January 1974 and 31 December 2010. The blue/light grey dots indicate that run-off is generated by snowpack melting combined with rainfall events. The red/dark grey dots indicate that run-off is generated by rainfall events alone. The 95th percentile of $RU2_e$ represented in the graphs is the limit above which the days are considered as favourable to floods. A flooding flow rate of $300 \text{ m}^3 \text{ s}^{-1}$ is considered. [Colour figure can be viewed at wileyonlinelibrary.com].

temperature (the 5th percentile of the daily minimum temperature, in agreement with Van de Vyver (2012)) are non-significant everywhere but show a positive signal in the Ardennes (see Figure S3(c)). Extreme maximum temperature in winter (the 95th percentile of the daily maximum temperature, in agreement with Van de Vyver (2012)) also exhibits non-significant trends but show a homogeneous positive signal between $+0.6$ and $+1.4^\circ \text{C} (52 \text{ years})^{-1}$ in the Ardennes (see Figure S3(d)). The absence of a clear trend in winter temperature is also highlighted by Van de Vyver (2012), who found a non-statistically significant linear growth in winter extremes for the period 1952/1953–2009 over Belgium. In addition, similar results are found using the MAR-NCEP1 outputs (see Figure S5). However, the trends computed from MAR-ERA-20C are positive, larger and significant in any part of Belgium (see Figure S10). It is easily explained by the fact that air temperature is not assimilated in ERA-20C (Poli *et al.*, 2013), which overestimates the global warming in Europe.

Regarding *precipitation* changes in winter, no significant trend is found in the Ourthe catchment. However, Figure 4(a) and (b) show a negative signal in both total rainfall and particularly snowfall valued at $-10 \text{ mm} (52 \text{ years})^{-2}$ at least. Figure 4(c) and (d) display a positive signal in both extreme precipitation events intensity (trend in the 95th percentile of the daily precipitation amount in winter) and frequency (trend in the number of days with daily precipitation amount larger than its 95th percentile computed over 1960–1991) up to $+2 \text{ mm years}^{-2}$ and up to $+2 \text{ events years}^{-2}$, respectively according to MAR-ERA. The trends computed using the MAR-ERA-20C outputs are similar but larger and thus statistically significant in some parts of the Ourthe catchment (see Figure S11). MAR-NCEP1 simulates

opposite trends in comparison with MAR-ERA and MAR-ERA-20C (see Figure S6). Therefore comparison with previous studies covering Belgium is needed to check the relevance of our results. Most of these studies identify an increase in the winter precipitation amount, along with an increase in both intensity and frequency of extreme precipitation events during winter (Gellens, 2000; Vaes *et al.*, 2002; De Jongh *et al.*, 2006; Ntegeka and Willems, 2008; Willems, 2013a, 2013b; Royal Meteorological Institute of Belgium (RMI), 2015). However, because the significance of these trends depends both on the length of the time series and on the duration of the precipitation events, the aforementioned studies conclude that no statistically robust trend can be acknowledged regarding precipitation in Belgium. Despite the absence of significant long-term trends, these studies identify multidecadal oscillations characterized by drier periods in the 1900s, around 1920 and in the mid-1970s (De Jongh *et al.*, 2006; Willems, 2013a, 2013b) alternating with wetter periods in the 1910–1920s, the 1950–1960s and in the 1990–2000s (Ntegeka and Willems, 2008; Willems, 2013a, 2013b).

By analysing the evolution of the seasonal *snow* cover in winter, a significant decreasing trend is identified in all three MAR simulations. According to MAR-ERA, the maximum of daily SH significantly decreases by 5 to $10 \text{ cm} (52 \text{ years})^{-1}$ in the Ardennes, and by more than $15 \text{ cm} (52 \text{ years})^{-1}$ in the Hautes Fagnes Plateau (see Figure 5(a)). Considering snow days as days with a snow cover of at least 5 cm of thickness, in agreement with Erpicum *et al.*, (1991), the number of snow days significantly decreases up to $-15 \text{ days} (52 \text{ years})^{-1}$ (see Figure 5(b)). The onset of the snow cover season (the first day of the year with a snow cover height of at least 1 cm) is delayed up to $+60 \text{ days} (52 \text{ years})^{-1}$ in some parts of

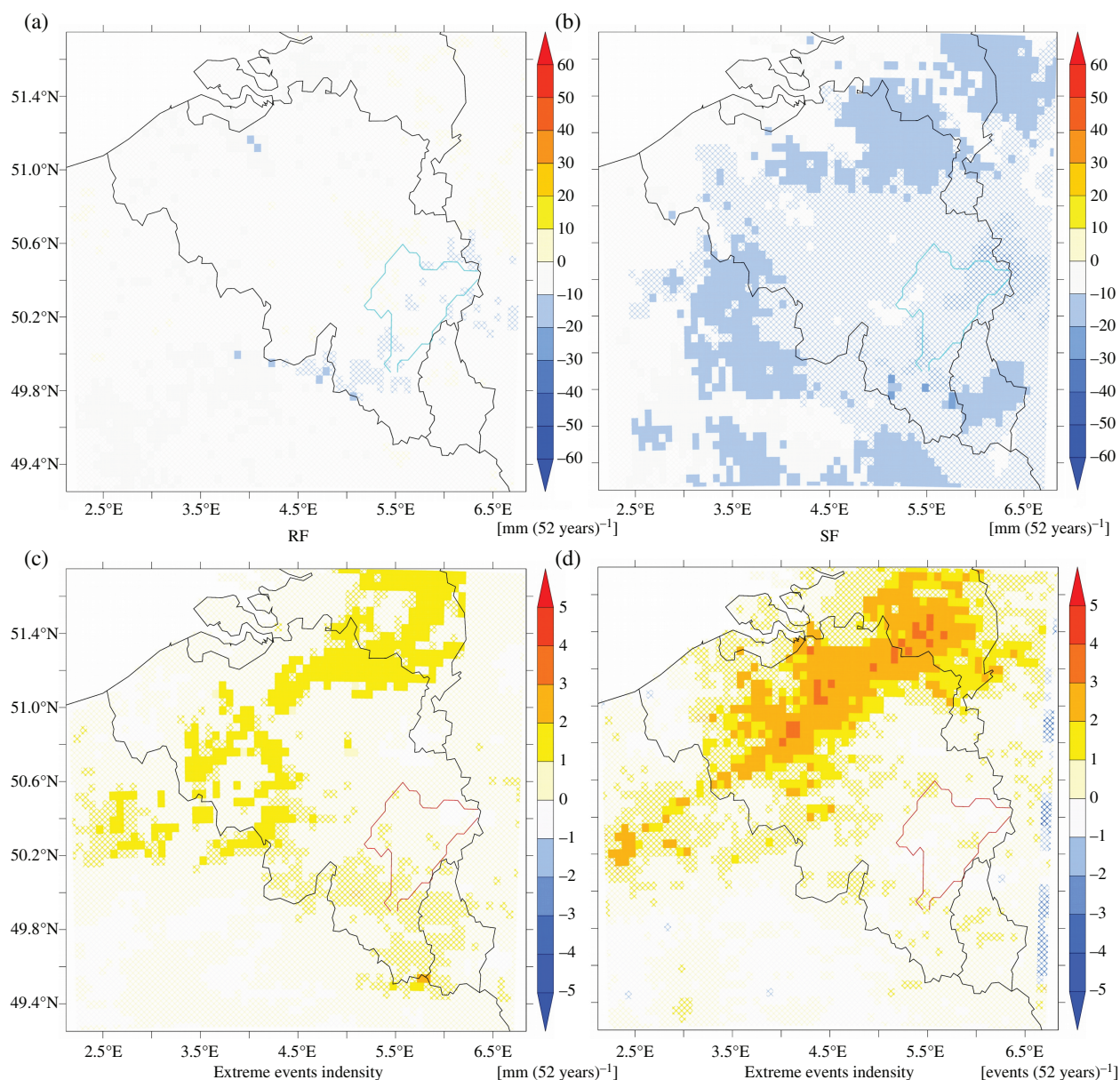


Figure 4. Trends computed from MAR-ERA over 1959–2010 in (a) the rainfall amount (RF) in winter (DJF), (b) the snowfall amount (SF) in winter, (c) the intensity of extreme precipitation events in winter (trend in the 95th percentile of daily precipitation amount in winter), and (d) the frequency of extreme precipitation events in winter (trend in the number of days in winter with daily precipitation amount larger than its 95th percentile computed over 1961–1990). Filled pixels indicate the places where the trends are statistically significant. [Colour figure can be viewed at wileyonlinelibrary.com].

the Ourthe catchment (see Figure 5(c)). By contrast, the end of the snow cover season (the last day of the year with a snow cover height of at least 1 cm) is shortened up to -60 days $(52 \text{ years})^{-1}$ only in a small part of the Ourthe catchment area (see Figure 5(d)). Therefore, the duration of the snow cover season seems to shorten throughout the last 52 years. Similar results are found using the outputs of MAR-NCEP1 (see Figure S7) and MAR-ERA-20C (see Figure S12). These results are in agreement with studies carried out in other regions in Europe, such as the low- and medium-elevated parts of the Alps (e.g. Durand *et al.*, 2009; Valt and Cianfarra, 2010; Beniston, 2012b), or the lowland and coastal regions of Norway (Skaugen *et al.*, 2012; Dyrørdal *et al.*, 2013).

Changes in the winter atmospheric regional circulation partly explains the trends found in our results. In Western Europe, these circulation changes are often assessed by analysing changes in the North Atlantic Oscillation (NAO) Index (e.g. Hurrell, 1995; Slonosky and Yiou, 2001; Philipp *et al.*, 2007). The NAO Index is defined as the difference of the atmospheric surface pressure between the Azores High and the Icelandic Low, which corresponds to the normalized difference of sea level surface pressure measured in Lisbon (Portugal) and in Stykkisholmur/Reykjavik (Iceland). In the wintertime, the NAO Index is known for its high influence on the weather of Western Europe and for its high day-to-day and year-to-year variability. For instance, negative NAO

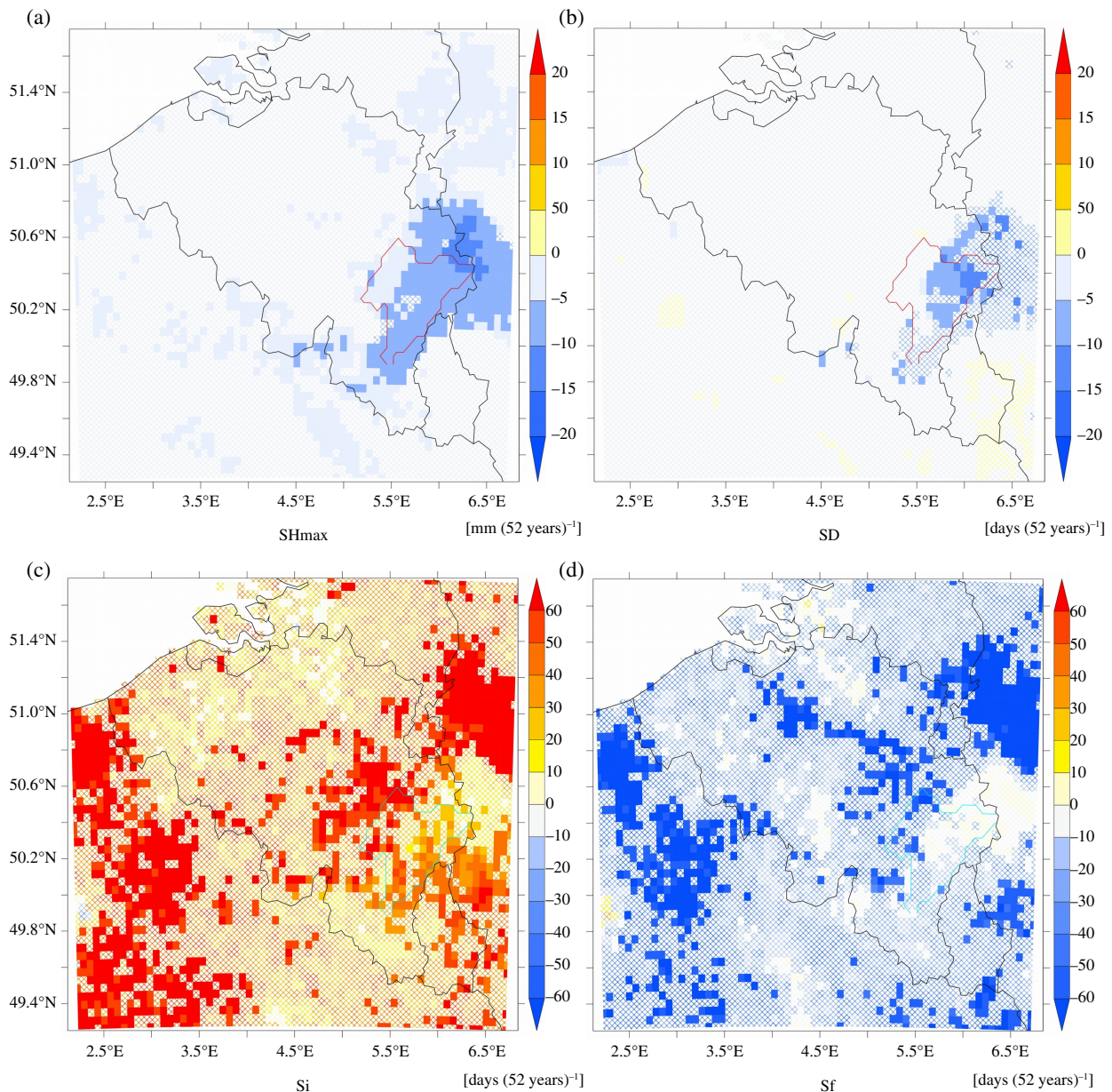


Figure 5. Trends computed from MAR-ERA over 1959–2010 in (a) the maximum daily snow depth (SD) in winter (DJF), (b) the number of days per winter with a snow cover of at least 5 cm of thickness (SD), (c) the first day of the year with a snow depth of at least 1 cm (Si), and (d) the last day of the year with snow depth of at least 1 cm (Sf). Filled pixels indicate the places where trends are statistically significant. [Colour figure can be viewed at wileyonlinelibrary.com].

Index values usually correspond to cold and snowy winters in Belgium such as the winter 2009–2010 (Cattiaux *et al.*, 2010), while positive NAO Index values are associated with mild and rainy winters.

As outlined by Van de Vyver (2012), the absence of clear trends in the winter temperatures of Belgium could be due to the perturbation of the warming signal by the high variability of the regional atmospheric circulation in winter, assessed from the variation of the NAO Index in winter (see Figure 6(b)).

Furthermore, as the decrease in snowfall does not bring more rainfall, it suggests that the (non-significant) increasing mean winter temperature would not be responsible for this declining snowfall trend and the resulting decrease

in snow accumulation. Thus, changes in the winter atmospheric regional circulation explain in part this decline of both precipitation amount and snow accumulation. Figures 6(a), S8(a) and S13(a) show that whatever the reanalysis forcing the MAR model, an increasing trend in the daily mean surface pressure in winter of about 1 to 3 hPa (52 years)⁻¹ is found, suggesting an increase in the occurrence of anticyclonic conditions in winter over the Belgian territory. In addition, Figure 6(b) shows a significant long-term increasing trend in the NAO Index during winter over the studied period, suggesting a strengthening in the Azores High. These more anticyclonic-like conditions in winter are likely to cause the decrease in total precipitation amount including snowfall, leading thus to

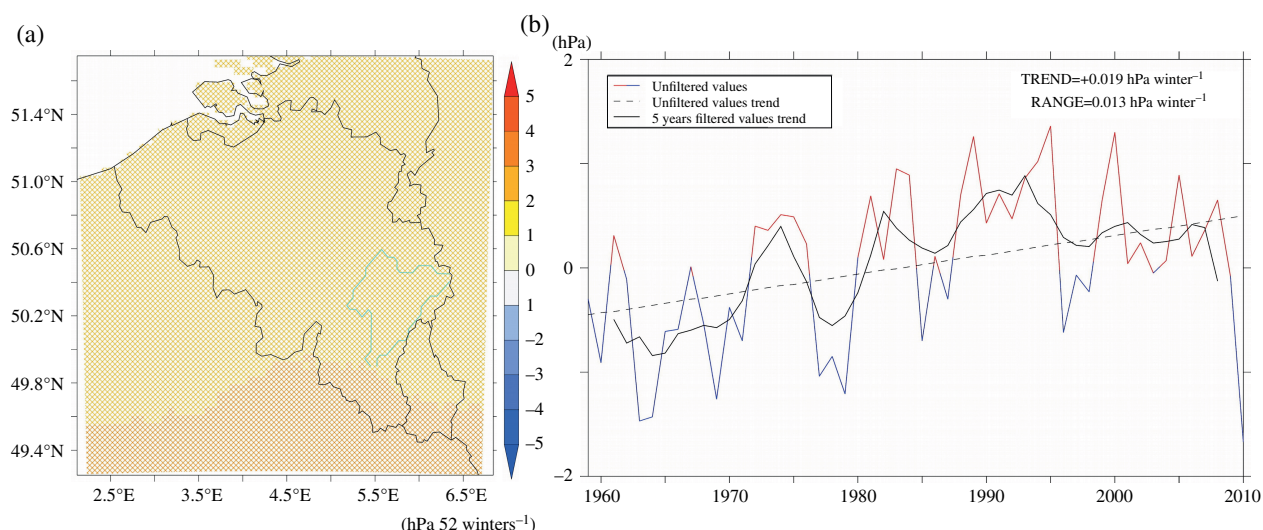


Figure 6. Trends over 1959–2010 of (a) the mean surface pressure in winter (DJF) computed from MAR-ERA, and (b) the mean NAO index recorded in winter. Filled pixels indicate the places where the trends are statistically significant. [Colour figure can be viewed at wileyonlinelibrary.com].

reduced snow accumulation. However, the MAR integration domain used in this study is not large enough to further investigate hypothetical circulation changes over Western Europe.

However, this decrease in the precipitation amount over the studied period may also depend on the period selected for conducting the trend analysis as the climate of Belgium is affected by multidecadal oscillations (e.g. Ntegeka and Willems, 2008; Willems, 2013a; 2013b).

Nonetheless, it should be noted that the decreasing trends in both precipitation amount and snow accumulation simulated by MAR in winter are in part biased by the inability of MAR to properly simulate convective precipitation (see Section 3.1.). Indeed, the MAR simulates an increase in convective precipitation during winter especially if MAR is forced by NCEP/NCAR-v1 (see Figure S9(b)). Therefore, the decrease in precipitation in winter simulated by MAR could be resulting from an inadequate convective scheme in the model, given that MAR is a hydrostatic model running at a resolution of 5 km. In addition, the raw reanalyses and previous studies using weather station observations have opposite conclusions.

In any case, the increase in convective precipitation in winter remains to be explained. As this increase is the greatest in coastal regions (see Figure S9(b)), warmer sea surface temperatures could have enhanced instability of the perturbations which usually cross the country from north-west to south-east and thereby bring convective precipitation. The North Sea is indeed acknowledged to be one of the seas belonging to the north-east Atlantic region which have recorded the fastest warming over the last 25 years (Joyce, 2006; Sherman *et al.*, 2007).

Lastly, it should also be noticed that the signal of the trends in both the intensity and the frequency of extreme precipitation events is not clear. The trends obtained from the MAR-NCEP1 outputs (negative signal) more particularly contrast with the signal obtained from the MAR-ERA and the MAR-ERA-20C outputs, as well as

with conclusions of previous studies. It brings to light that the reanalysis data (in particular NCEP/NCAR-v1) may be too coarse to allow to MAR to capture extreme precipitation events in an accurate way because we directly force the MAR model with the reanalysis.

3.2.2. Changes in conditions favourable to floods

Because more than 70% of the floods in the Ourthe River occurred in winter, changes in conditions favourable to floods are investigated by focusing on this season. Days favourable to floods (as defined in Section 3.1.) computed from the MAR-ERA results are summed for each winter of 1959–2010 [see total flood days (TFD) in Figure 7(c)]. A distinction is also made between days favourable to floods resulting from snowpack melting combined with precipitation [see (snowmelt-induced flood days SFD) in Figure 7(a)], and days favourable to floods from precipitation only [see precipitation-induced flood days (PFD) in Figure 7(b)]. Then, for each of these time series, trends and their significance are computed following the method described in Appendix S1. Finally, a 5-year running mean is applied to the time series. The same work is also repeated for the MAR-NCEP1 and MAR-ERA-20C outputs and is available in Figures S15 and S16.

First of all, long-term linear trends show that MAR-ERA exhibits a significant decreasing trend in SFD of -0.100 days/winter (see Figure 7(a)). Results obtained using MAR-NCEP1 and MAR-ERA-20C confirm the decreasing trend in SFD. However, regarding PFD, the results contrast following the reanalysis forcing the MAR model because MAR-ERA shows no trend (see Figure 7(b)), MAR-NCEP1 shows a significant decreasing trend and MAR-ERA-20C computes a non-significant increasing trend. As a result, regarding TFD, MAR-ERA shows a non-significant decreasing trend valued at -0.084 days/winter (see Figure 7(c)), MAR-NCEP1 exhibits a significant decrease in TFD over the studied

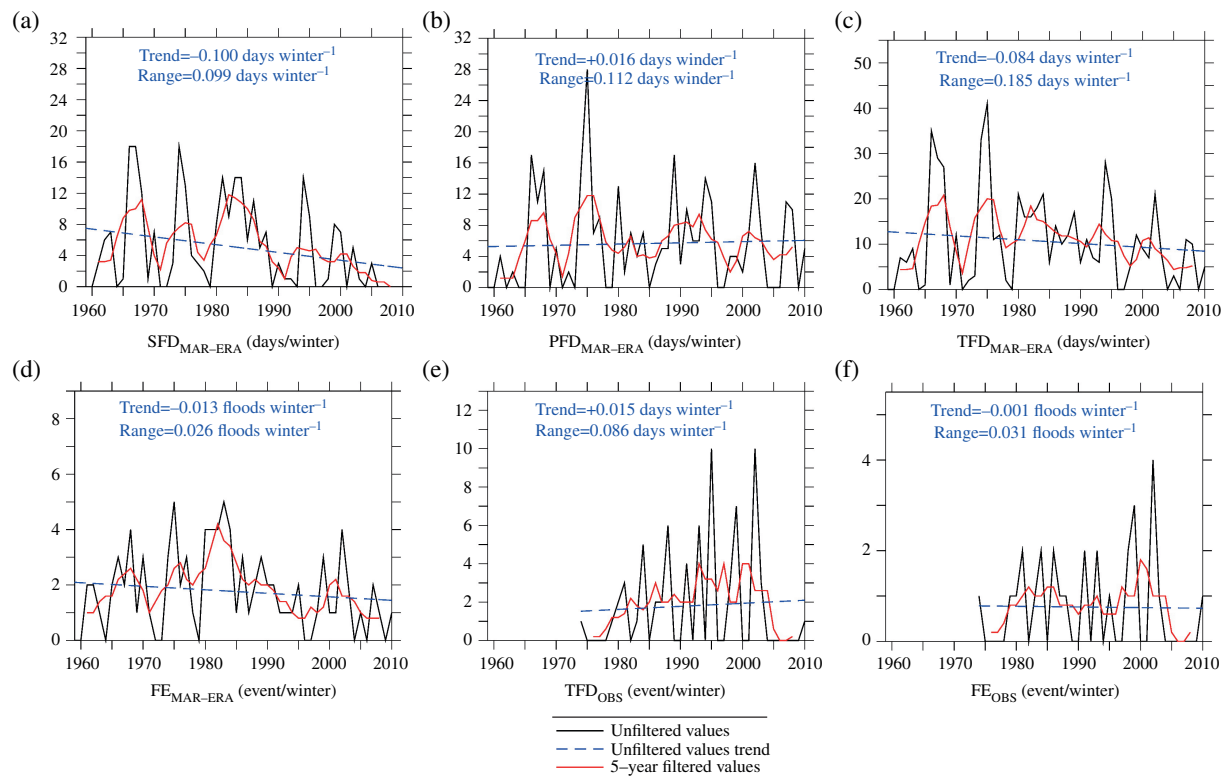


Figure 7. Unfiltered values, 5-year filtered values and trends computed from MAR-ERA over 1959–2010 of (a) the number of days per winter (DJF) favourable to floods due to snowpack melting combined with rainfall events ($SFD_{MAR-ERA}$), (b) the number of days per winter favourable to floods due to rainfall events alone ($PFD_{MAR-ERA}$), (c) the total number of days per winter favourable to floods ($TFD_{MAR-ERA}$), (d) the number of flood events favourable to floods per winter ($FE_{MAR-ERA}$). Unfiltered values, 5-year filtered values and trends computed from observations in Sauheid are also displayed with (e) the number of observed flood days per winter (TFD_{OBS}), and (f) the number of observed flood events per winter (FE_{OBS}). [Colour figure can be viewed at wileyonlinelibrary.com].

period while MAR-ERA-20C exhibits no trend. Observations of effective flood days (TFD_{OBS}) in Sauheid, based on flow rate measurements between 1974 and 2010, show no trend (see Figure 7(e)). In terms of (FE), only MAR-NCEP1 exhibits a significant trend valued at -0.042 floods/winter (see Figure S15(d)) while no significant trend can be found in the observations (see Figure 7(f)). However, as MAR underestimates convective precipitation, especially if the model is forced by NCEP/NCAR-v1, extreme run-off events generated by this kind of precipitation are probably underestimated. Therefore, the trends in conditions favourable to floods computed from the MAR results are likely to be slightly overestimated, given that the ratio of convective precipitation events versus stratiform precipitation has been increasing since 1959 (see Figure S9(b)).

On the basis of FE observed over 1974–2010 in Sauheid, two periods of higher flood occurrence are identified in the 1980s, in agreement with Pauquet and Petit (1993), and also in the 2000s (see the 5-year running mean (filtered) values in Figure 7(f)). The period of higher flood occurrence in the 1980s seems to be partly induced by an increase in hydroclimatic conditions favourable to floods, as a peak is also simulated by all three MAR configurations during this period in SFD (see the 5-year running mean values in Figure 7(a)). The higher observed flood occurrence in the 2000s could also be partly related to

an increase in favourable hydroclimatic conditions, as all three MAR simulations, and more particularly MAR-ERA and MAR-ERA-20C, show a new peak in simulated SFD (see the 5-year running mean values in Figure 7(a)) and especially in simulated PFD (see the 5-year running mean values in Figure 7(b)). Although observations about FE frequency are available only after 1974, we identify another peak in the 1960s in SFD (see the 5-year running mean values in Figure 7(a)), PFD (see the 5-year running mean values in Figure 7(b)). Since this peak coincides with the higher flood frequency period of the 1960s observed by Pauquet and Petit (1993), we assume that this period of higher floods frequency was also induced by favourable hydroclimatic conditions. In addition, previous studies identified the 1960s as a wetter period (Ntegeka and Willems, 2008; Willems, 2013a, 2013b).

4. Conclusions and prospects

The main objectives of this paper were first to determine whether significant trends in snow cover evolution exist over the period 1959–2010 in the Belgian Ardennes. Secondly, since most of the floods that take place in the Ardennes rivers occur in winter as a result of rapid snowpack melting and/or heavy rainfall events, this paper also aimed to assess the changes in hydroclimatic conditions

that could generate floods in the Ourthe, a river of the Ardennes. The recent changes in climate were reconstructed using the RCM MAR forced by several reanalysis datasets: ERA-Interim (1979–2014) completed by ERA40 (1958–1978), NCEP/NCAR-v1 (1958–2014) and ERA-20C (1958–2010).

Trends over the period 1959–2010 were computed from the MAR outputs by focusing on the winter months. In the Ourthe catchment area, neither MAR-ERA nor MAR-NCEP1 show any significant trends whether in winter mean temperature or in extreme (P5 and P95) temperature. Regarding precipitation in winter, MAR-NCEP1 and MAR-ERA-20C show a significant decrease in large part of the Ourthe catchment ranging from -10 and -40 mm (52 years) $^{-2}$. Trends in both intensity and frequency of extreme precipitation events show a positive signal with MAR-ERA and with MAR-ERA-20C reaching $+4$ mm (52 years) $^{-2}$ and $+4$ extreme events (52 years) $^{-2}$ in some parts of the Ourthe catchment. All MAR configurations exhibit a significant decrease in the maximum SH reached during winter in the Ardennes over 1959–2010, with trends valued between -5 and -15 cm (52 years) $^{-1}$ in the Ourthe catchment. The number of days with at least 5 cm of snow accumulation is reduced up to -15 days (52 years) $^{-1}$, and the beginning of the snow season is delayed up to -60 days (52 years) $^{-1}$ in the highest parts of the Ourthe catchment. Because the NAO is known to highly impact the weather of European winters, the large interannual variability of the NAO Index explains the absence of significant trends in winter temperature in Belgium in comparison with the observed global warming. The general increase of the NAO index observed since 1959 suggests a higher occurrence of anticyclonic conditions, which could be responsible for the decrease in winter precipitation and consequently for the decrease in snow accumulation. However, the multidecadal oscillations in Belgian precipitation could also explain the absence of clear trend in precipitation amount and extremes.

Regarding hydroclimatic conditions favouring floods in winter in the Ourthe catchment, the number of days favourable to floods due to snow melting combined with rainfall events is decreasing over 1959–2010 whatever the MAR simulations. This results from a decrease in snow accumulation in winter and from a shortening of the snow season over the studied period. Regarding the number of days favourable to floods due to intense rainfall alone, the signal depends on the reanalysis used as forcing. It is due to the fact that the signal and the pattern of the trends in both the intensity and the frequency of extreme precipitation events differ depending on the MAR forcing. In a warmer climate, we could assume that the current trend of decreasing floods due to abrupt melting of the snowpack should accelerate. Yet, this decreasing trend favouring less floods in Belgium could be compensated by an increase in the occurrence of extreme rainfall events and by an increase in precipitation amount in autumn and winter as highlighted by Baguis *et al.* (2010), Willems (2013c) or Tabari *et al.* (2015). Similar simulations using global models from the CMIP5 (Coupled Model Intercomparison Project Phase 5)

database as forcing should help to better predict potential changes at the end of this century. It would allow to provide dynamic downscaling results to enhance the statistical downscaling findings (Tabari *et al.*, 2015).

This study brings to light the differences and the limits of the reanalyses used to force the MAR model. Indeed, the comparison of the model results with observations over 2008–2014 showed that MAR-ERA produces slightly better results than MAR-ERA-20C and particularly MAR-NCEP1. In addition, the trend study shows that the results differ according to the reanalysis forcing MAR. It also shows that given the unclear signal for the impact of extreme precipitation evolution on conditions favouring floods, the reanalysis data is maybe too coarse to allow to MAR to capture those events in an accurate way. A quick comparison with the literature allows to determine which reanalysis forcing the MAR model is the best to study each variables.

- When used as MAR forcing, *NCEP/NCAR-v1* produces the worst results when compared with observations, especially in terms of precipitation amount measurements and consequently floods dates. The trend study shows that MAR forced by this reanalysis produces trends in opposition with previous studies and the two other MAR simulations regarding the wintertime extreme events intensity and frequency, as well as conditions favourable to floods generating by intense rainfall events. However, MAR-NCEP1 produces consistent results regarding trends in temperature, snow accumulation and snow season. The weaknesses of MAR-NCEP1 could be explained by the low-spatial resolution of the NCEP/NCAR-v1 reanalysis as the MAR model is directly forced by this reanalysis. Therefore, using MAR-NCEP1 simulations performed at an intermediate resolution as forcing of the 5 km MAR would possibly improve the results. However, it should be noticed that Kay *et al.* (2015) found that the direct use of data from the intermediate resolution is generally better for hydrological modelling.
- *ERA-20C* produces good results when compared with observations. However, the trends in winter temperature computed from MAR-ERA-20C are overestimated as well as the trends in snow season. Despite better spatial resolution than NCEP/NCAR-v1, air temperature data are not assimilated in ERA-20C which potentially explains this bias in temperature. However, MAR-ERA-20C produces consistent results regarding trends in extreme precipitations events, trends in snow accumulation and consequently trends in hydroclimatic conditions favourable to floods.
- *ERA-Interim* produces the best results when compared with observations and conclusions of previous studies. It is due to a better data assimilation system and a better spatial resolution in comparison with NCEP/NCAR-v1 or ERA-20C.

Finally, this study highlights the strengths and the weaknesses of the MAR model running over a temperate region.

Regardless of the reanalysis used as forcing, MAR showed its ability to reconstruct the variability of the Belgian climate, especially the daily variability of temperature and snow depth. However, MAR underestimates temperature in winter and conversely, it overestimates temperature in summer. MAR also underestimates snow depth. Moreover, MAR shows difficulties to simulate precipitation in terms of daily variability and amount. Last of all, although the model allows to reconstruct hydroclimatic conditions favouring floods in the Ourthe catchment, it largely overestimates the number of days which could have generated floods over 1974–2010. Therefore, in the view of these discrepancies, improvements in the model are required such as:

- The re-parametrization of the fresh snowfall density would correct the negative bias in snow accumulation. Indeed, we used a fresh snowfall density of 300 kg m^{-3} which is typical in polar regions while lower values are more relevant in temperate climates ($\sim 100 \text{ kg m}^{-3}$).
- The improvement of the convective scheme parametrized in MAR would partly correct precipitation biases as well as temperature biases in summer. However, it should be noticed that studies using convection-permitting RCM have shown that, while some aspects of precipitation are improved, some are not improved or can even be worse (e.g. Kendon *et al.*, 2012; Chan *et al.*, 2013).
- The improvement of the surface and soil properties of MAR would also partly correct precipitation and temperature biases because test simulations demonstrate that wetter soils allow to correct the temperature biases in summer. Currently, only the first metre of soil is resolved in MAR. A deeper soil (10 m) could improve the results. In addition, the coupling of MAR with a vegetation model would allow MAR to have dynamic surface properties.
- The two-way coupling between MAR and a hydrological model would allow to improve floods detection and simulation by taking into account mechanisms which are too simplified or not represented in the RCMs. Moreover, it would possibly reduce the temperature and precipitation biases by enhancing energy and water fluxes between the surface/soil and the atmosphere (e.g. Larsen *et al.*, 2016; Wagner *et al.*, 2016).

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Supporting information

The following supporting information is available as part of the online article:

Appendix S1: Linear regression and uncertainty range for the 95% confidence interval of Snedecor.

Figure S1: The MAR model integration domain and orography.

Figure S2: Taylor diagrams showing relative differences between MAR-ERA, MAR-NCEP1, MAR-ERA-20C, ERA-Interim, NCEP/NCAR-v1 and ERA-20C in comparison with seasonal observations of daily mean temperature and daily precipitation amount over 2008–2014.

Figure S3: Trends computed from MAR-ERA over 1959–2010 in the mean winter temperature, the yearly mean temperature, the extreme minimum temperature in winter (DJF) and the extreme maximum temperature in winter, over Belgium.

Figure S4: Trends computed from MAR-ERA over 1959–2010 in the percentage of stratiform and convective precipitation amount in winter (DJF), over Belgium.

Figure S5: Same as Figure S3 but for MAR-NCEP1.

Figure S6: Trends computed from MAR-NCEP1 over 1959–2010 in the rainfall and snow fall amount in winter (DJF), the intensity of extreme precipitation events in winter and the frequency of extreme precipitation events in winter, over Belgium.

Figure S7: Trends computed from MAR-NCEP1 over 1959–2010 in the maximum daily snow height in winter (DJF), the number of days with a snow cover of at least 5 cm of thickness in winter, the first day of the year with a snow cover of at least 1 cm and the last day of the year with a snow cover of at least 1 cm, over Belgium.

Figure S8: Trends computed from the MAR-NCEP1 outputs over 1959–2010 in the mean surface pressure in winter (DJF) over Belgium and trend in the mean NAO Index recorded in winter over the same period.

Figure S9: Same as Figure S4 but for MAR-NCEP1.

Figure S10: Same as Figure S3 but for MAR-ERA-20C.

Figure S11: Same as Figure S6 but for MAR-ERA-20C.

Figure S12: Same as Figure S7 but for MAR-ERA-20C.

Figure S13: Same as Figure S8 but for MAR-ERA-20C.

Figure S14: Same as Figure S4 but for MAR-ERA-20C.

Figure S15: Unfiltered values, 5-year filtered values and trends computed from the wintertime MAR-NCEP1 results over 1959–2010 of the number of days favourable to floods due to snowpack melting associated with intense

rainfall events, the number of days favourable to floods due to intense rainfall events, the total number of days favourable to floods, the number of events favourable to floods, in the Ourthe catchment. Flood days and events observed in Sauheid in winter are also displayed.

Figure S16: Same as Figure S15 but for MAR-ERA-20C.

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