Climate impact on floods: changes in high flows in Sweden in the past and the future (1911-2100)

Berit Arheimer and Göran Lindström

Swedish Meteorological and Hydrological Institute, 601 76 Norrköping, Sweden

Berit.Arheimer@smhi.se

Abstract: There is an ongoing discussion whether floods occur more frequently today than in the past, and whether they will increase in number and magnitude in the future. To explore this issue in Sweden, we merged observed time series for the past century from 69 gauging sites throughout the country (450 000 km^2) with high-resolution dynamic model projections of the upcoming century. The results show that the changes in annual maximum daily flows in Sweden oscillate between dry and wet periods but exhibit no significant trend over the past 100 years. Temperature was found to be the strongest climate driver of changes in river high flows, which are related primarily to snowmelt in Sweden. Annual daily high flows may decrease by on average -1% per decade in the future, mainly due to lower peaks from snowmelt in the spring (-2% per decade) as a result of higher temperatures and a shorter snow season. In contrast, autumn flows may increase by +3% per decade due to more intense rainfall. This indicates a shift in flood-generating processes in the future, with greater influence of rain-fed floods. Changes in climate may have a more significant impact on some specific rivers than on the average for the whole country. Our results suggest that the temporal pattern in future daily high flow in some catchments will shift in time, with spring floods in the north-central part of Sweden occurring about one month earlier than today. High flows in the southern part of the country may become more frequent. Moreover, the current boundary between snow-driven floods in north-central Sweden and rain-driven floods in the south may move toward higher latitudes due to less snow accumulation in the south and at low altitudes. The findings also indicate a tendency in observations toward the modeled projections for timing of daily high flows over the last 25 years. Uncertainties related to both the observed data and the complex model chain of climate impact assessments in hydrology are discussed.

Keywords: long-term records, S-HYPE, climate change, catchment modeling, river flow, trend analysis

1. Introduction

Numerous severe floods have been reported globally in recent years, and there is growing concern that flooding will become more frequent and extreme due to climate change. Generally, a warmer atmosphere can hold more water vapor, in effect leading to a growing potential for intense precipitation that can cause floods (Huntington, 2006). Some scientists have argued that the observed changes in climate (e.g., increases in precipitation intensity) are already influencing river floods (e.g., Kundzewicz et al., 2007; Bates et al., 2008). However, there are methodological problems associated with detection of changes in floods, and large uncertainties arise when exploring trends in both past and future high flows.

Changes in river flood regimes are traditionally analyzed using statistical approaches and observed data (e.g., Lindström and Alexandersson, 2004; Stahl et al., 2010; Schmocker-Fackel and Naef, 2010) or process-based numerical modeling and a scenario approach (e.g., Dankers and Feyen, L., 2008; Arheimer et al., 2012; Bergström et al., 2012). Both these strategies have potential advantages but also many challenges, as discussed by Hall et al. (2014). The two fundamental problems connected with climate impact assessments can be summarized as follows: (1) observed time series can include natural long-term cycles that might be induced by climatic oscillations or persistent memory of hydrological processes (Markonis and Koutsoyiannis, 2012; Montanari, 2012), which will render all statistical trend analyses very sensitive to the period chosen for the study; (2) global climate models (GCMs) do not correspond to the observed climatology (Murphy et al., 2007), and uncertainties arise in each step of the model chain in hydrological impact assessments (Bosshard et al., 2013; Donnelly et al. 2014). Much effort has been made over the last decade to address these problems by finding more robust methods for analyzing trends and scenario models (see full review in Hall et al., 2014).

Most studies in the literature relate changes in climate to mean annual flow, whereas few concern the impact of such changes on high flows or consider specific drivers. One way to understand changes in flood-generating processes is to analyze seasonality. Some of the main driving processes (e.g., cyclonic precipitation, convective precipitation, and snowmelt events) are highly seasonal, and thus studying flood occurrence within a year may provide clues regarding flood drivers and changes in those factors (e.g., Parajka et al., 2009; Petrow and Merz, 2009; Kormann et al., 2014).

Hall et al. (2014) argue that future work should exploit, extend, and combine the strengths of both flow record analysis and the scenario approach. The present study concurs with the idea of merging analysis of long time series from the past with dynamic scenario modeling of the future. Climate change detection should be based on good quality data from observation networks of rivers with near-natural conditions (e.g., Lindström and Bergström, 2004; Hannah et al. 2010), and time series of more than 50-60 years are recommended to account for natural variability (Yue et al., 2012, Chen and Grasby, 2009).

Accordingly, we used time series spanning 100 years (1911-2010) from 69 gauged unregulated rivers to examine recorded changes in flood frequency and magnitude. Modeling of the future was performed according to the typical impact modeling chain "emission scenario - global climate model - regional downscaling - bias correction - hydrological model - flood frequency analysis", and this was done using the Swedish national hydrological model system S-HYPE with observed climatology and two 100-year (2000-2100) climate model projections. An overlapping period of 50 years was applied to check agreement between observed and modeled trends in high flows. The following questions were addressed:

(i) What changes have occurred in daily high flows in Sweden during the last century, and what changes can be expected over the next hundred years?

(ii) What climate drivers cause such changes?

(iii) How will the flood regime and dominating flood-generating processes change in the future?

2. Data and methods

2.1 Landscape characteristics and high flows observed in the past

Sweden is located in northern Europe and has a surface area of about 450 000 km^2. Approximately 65% of the country is covered by forest, but there are major agricultural areas in the south. Sweden is bordered by mountains to the west and a long coastline to the south and east, and hence the country is drained by a large number of rivers that have their sources in the west and run eastwards to the Baltic Sea and southwest to the North Atlantic. Most of the rivers are regulated, and around 50% of the electricity in Sweden comes from hydropower. To detect general tendencies in flood change, we aggregated results from analyses of long-term records and scenario modeling to the four regions (Fig. 1) defined by Lindström and Alexandersson (2004). The river basins in these regions show similarities in climate and morphology, but also represent the catchments of the marine basins. Sixty-nine gauges with long records and very little or no upstream regulation in the catchment were chosen from the national water archive to represent the four regions (Fig. 1).

2.2 Model approach to the past and the future

Water discharge and hydrometeorological time series for the past and the future were extracted from the Swedish multi-basin model system S-HYPE (Strömqvist et al., 2012, Arheimer and Lindström, 2013), which covers more than 450 000 km^2 and produces daily values for hydrological variables in 37 000 catchments from 1961 onwards. This system is based on the process-derived and semi-distributed Hydrological Predictions for the Environment (HYPE) code (Lindström et al., 2010), and it comprises the Swedish landmass including transboundary river basins. The first S-HYPE was launched in 2008, but the system is continuously being improved and a new version is released every second year. Observations from 400 gauging sites are available for model evaluation of daily water discharge. The S-HYPE version from 2010 was used in the present study.

We forced the S-HYPE model with daily precipitation and temperature data, using national grids of 4 km based on observations and climate model results, respectively. The grid based on daily observations was produced using optimal interpolation of data from some 800 meteorological stations, considering variables such as altitude, wind speed and direction, and slopes (Johansson, 2002). To study floods, gridded values were transformed to each subbasin for the period 1961-2010 to force the S-HYPE model.

For climate model results, we used two grids based on different general circulation models (GCMs): HadCM3Q0 (Johns et al., 2003; Collins et al., 2006) and ECHAM5r3 (Roeckner et al., 2006). The projections were chosen to represent different signals concerning future climate change. In the ensemble of 16 climate projections studied by Kjellström et al. (2011), the Hadley projection is among those with the largest future temperature increase in Scandinavia, and the Echam projection represents those with low to medium increase. Bosshard et al. (2014) considered all possible selections of two projections from this ensemble and noted that the chosen projections spanned a larger uncertainty range than at least 70% of the other combinations. Both projections simulated effects of the emission scenario A1B (Nakićenović et al., 2000) and the GCM results were dynamically downscaled to 50 km using the RCA3 model (Samuelsson et al. 2011). Thereafter, daily surface temperatures (at 2 m) and precipitation were further downscaled to 4 km, and bias was corrected using the distribution-based scaling method (Yang et al. 2010) with reference data from the 4 km grid-based observations for 1981-2010. Finally, gridded values were transferred to each subbasin for the period 1961-2100 to force the S-HYPE model.

2.3 Quality check and analysis

The capacity of the model to predict annual maximum daily flows was tested at 157 gauging sites without regulation using S-HYPE version 2010. Model deviation for the calibration period and for an independent validation period of the same length was calculated using the forcing grid based on observations. Moreover, simulated trends in the various simulations were compared. Observed and modeled time series for 1961-2010 overlapped, and hence this 50-year period was used to check agreement between simulations. Observed and modeled results from the 69 river gauges were extracted and compared for different time slots. Simple linear regression was used as a trend test, because a previous study had shown no substantial discrepancy in results obtained by applying different trend tests to Swedish flood data (Lindström and Bergström, 2004). Statistical significance ($P = 0.05$) was estimated using the formula given by Yevjevich (1972, page 239).

To explore the spatial variability of climate change, the high-resolution results from the S-HYPE modeling were plotted as maps for two time windows (mean values for 2035-2065 and 2071-2100), showing estimated change for each climate projection. Furthermore, the annual distribution of daily high flows was plotted for the past and future in 15 selected catchments across the country to identify emergent patterns in seasonality.

To quantify temporal changes in annual high flows, we divided recorded values for the 69 gauges and modeled data from the 37 000 subbasins by the average value for the reference period (1961-1990) to obtain the relative anomalies at each site. These anomalies were then averaged separately for the country and each region to arrive at a relative change for each domain and each year. Frequency analysis was based on the proportion of gauging sites that exceeded the 10-year flood. The frequency was determined for each year in each region.

To relate climate drivers to flood changes, time series of temperature and precipitation data were extracted from the S-HYPE model for each subbasin and dataset (1961-2010 and 1961-2100). Also, these data were averaged for the country and each region based on site-specific annual anomalies compared to the long-term average for the reference period at each subbasin. Relative changes were considered for average and extreme precipitation, but absolute values were used for temperature (at 2 m).

To distinguish major long-term changes in the flood-generating mechanisms, seasonal changes in magnitude and frequency of high flows were analyzed by separating peaks occurring in March-June and July-February, respectively. In Sweden, spring peaks occurring in March to June along the south-to-north climate gradient are driven mainly by snowmelt, whereas autumn/winter peaks are primarily rain driven. Thus analyzing each group separately can provide information about any shift in hydrological regime and dominant processes that can cause high flows. We also investigated variation in timing of daily high flows in specific rivers in 15 selected catchments to assign changes to catchment-specific processes. In this assessment, the last 25 years, which were very mild, were highlighted to illustrate any shift toward the projected future.

Model results presented here were subjected to Gauss filtering, with a standard deviation corresponding to a moving average of 10 years, to distinguish between flood-rich and flood-poor periods in the long time series. The trend of the Gauss curves provides a clearer picture of the possible climate trend without the noise from single years. In addition, climate model results for single years should not be regarded as representative of specific years, because such models give long-term projections, not forecasts for individual years.

3. Results

The four hydroclimate regions in Sweden were analyzed both separately and combined using the 69 catchments and the S-HYPE model. However, this showed no clear difference in trends between regions, and therefore all results presented below apply to the entire country.

3.1 Observed annual maximum daily flows during the past

Over the last hundred years, the observed anomalies in annual maximum daily flow were normally within ±30% deviation from the mean of the reference period (Fig. 2). From the 1980s to 2010, the variability in flood frequency was less pronounced. One exception to this was the major flood event in 1995 involving at least a 10-year flood at most of the 69 gauging sites. This was linked to the very high spring flood, especially in the north, where previous maximum discharge records were exceeded by as much as 60% at some gauges. Spring floods normally corresponded to the annual high flow (cf. the two middle panels in Fig. 2), with a few exceptions, such as the autumn flood in 2000, which affected the central-southern parts of Sweden.

Considering the last hundred years, we found no obvious trends in magnitude of high flows in the observed time series, which was further confirmed by the statistical test (see Sect. 3.6). There was a slight decrease in flood frequency during this period, although in a shorter perspective it seems that autumn floods increased substantially over the last 30 years. It appears that 1970 was the turning point, and the summer and autumn floods in the 1920s were actually higher than in recent decades.

3.2 Model performance and comparison of trends in simulations

In the S-HYPE model (version 2010), the median absolute error was 15% for annual maximum daily flows at 157 gauging sites for both the calibration and the validation period (Fig. 3). Median underestimation was -0.7% for calibration but -3.5% for validation. The major outliers could be related to some missing lakes in this version of the S-HYPE model, as the model then overestimated high flows because the dampening effect of lakes was missing in the model setup.

Comparison of S-HYPE simulations using different forcing data revealed no statistically significant trends in observed or modeled high flows for the entire 50-year overlapping period (Table 1), although there was a small deviation between that period and the reference period, which was only 20 years shorter. Accordingly, the trend test detected no significant trends in shorter periods, except for the Hadley forcing, which showed statistically significant trends during the independent period and the reference period. In general, climate projections are not necessarily in phase with observed climate fluctuations, which was the case with the projections used in our study. This was also apparent for the longer time period of 50 years, for which the Echam forcing showed an opposite sign of slope compared to forcing with either Hadley or observed climate data.

The slope of the modeled time series using observed climate data was generally larger than the slopes of observed trends. This suggests that the S-HYPE model overestimates the sensitivity to changes in forcing data, or that there are compensating processes not included in the S-HYPE model (e.g., changes in land use, vegetation, or abstractions). The difference in slope may also be an artefact of bias in precipitation data, as discussed by Lindström and Alexandersson (2004) and Hellström and Lindström (2008). In four of the five time slices we examined, the S-HYPE model forced with observed climate data exhibited the same sign of slope as observed time-series of river flow. Again, it should be noted that none of these trend slopes were statistically significant (Table 1).

3.3 Future climate projections for Sweden

Figure 4 shows the large differences we obtained in spatial patterns of precipitation and temperature across Sweden when forcing S-HYPE with the two future projections. The results for the reference period (1981-2010) were similar for the two climate projections (Hadley and Echam), because precipitation and temperature were scaled against the same 4-km grid based on observations. However, considering future climate change, results provided by climate models differ greatly and can be conflicting, particularly regarding local conditions.

According to both projections in our study, the mean temperature will increase by 3-5 degrees in different parts of Sweden in the future. The Hadley model indicated a more rapid increase compared to the Echam model. The two models projected that average precipitation will increase by 100-400 mm per year depending on the geographical location, and the Hadley model indicated a faster and more marked increase.

The simulated change in average river flow varied ± 30% for different parts of Sweden. The model results based on the Echam forcing showed higher flow in the northern mountains and decreased flow in the rest of the country by the end of the century (Fig. 4). In contrast, Hadley forcing indicated increased river flow in all of northern Sweden and a decrease mainly in the south-eastern part of the country. This difference in river flow can be ascribed primarily to a combined effect of precipitation and evapotranspiration in the hydrological model. The precipitation emanated from the climate models, whereas the evapotranspiration in the HYPE model was calculated based on temperature values from the climate models. The large difference between the results of the climate projections implies considerable uncertainty in estimating future conditions.

There was ± 50% spatial variation in the future changes in mean high flow and the magnitude of the 10-year flood (Fig. 4), whereas for most of the country such divergence was only 15%. The estimated levels were highest for the northern part of the mountain range and south-western Sweden. The 10-year flood flows were lower for the mountains of Jämtland County, which is one of the areas with the most rich snowfall. There is a large spread in the results for the two projections, hence the findings regarding high flows on the local scale should be interpreted with caution. The Hadley forcing led to larger changes for the whole country, whereas Echam forcing indicated smaller changes compared to the reference period.

Our findings confirm that assessments of future climate change can differ markedly depending on the climate model that is applied, even if the same emission scenario is used. The two projections in our study were far from covering the full range of uncertainty, although a closer analysis shows that they did include most of the range of the ensemble of 16 climate projections used before, especially at the higher end of the extremes. The corresponding river flows calculated with S-HYPE were within the 25-75% range of the larger ensemble when using the HBV model (Bergström et al., 2012).

3.4 High flow in the future and climate drivers

Figure 5 shows that even though the forcing datasets were not in phase with each other or with the observations, similar trends for the future could be detected. Most substantial is the 5 degree rise in temperature by the end of the century in both projections. A strong increase in precipitation is also apparent, regarding both annual means and maximum daily levels. Similar trends can be seen in the observed data, although limiting the assessment to the past 50 years with a 10-year Gauss filter represents a rather short overlapping period for trend analysis. However, the strong trend in precipitation is not reflected in the river flow. It should also be noted that the temperature signal during the past 50 years in Fig. 5 is not representative of the 20th century as a whole (Lindström and Alexandersson, 2004).

Considering annual maximum daily flow in Fig. 5 reveals no trend over the past 50 years and a decreasing trend in the future. This can be explained by elevated temperature leading to lower spring peaks from snowmelt, caused by a shorter snow period and higher evapotranspiration. The results did not show any clear trend in 10-year flood frequency. Thus it seems that in Sweden, temperature is stronger than precipitation as a climate driver of river high flow, which illustrates that high flows are mainly related to snowmelt in this country.

3.5 Changes in the flood regime

The most substantial effect of changes in floods in Sweden was found when comparing the results of separate analyses of annual maximum daily flows occurring in the spring and in the rest of the year (mainly autumn). Figure 6 shows a significant decrease in magnitude of spring floods and a significant increase in autumn floods. For spring floods, using observed forcing data resulted in a weak trend, whereas the trend obtained using climate projections indicates 10-20% reduction by the end of the century compared to the 1970s.

For autumn floods the trend was in the opposite direction, with 10-20% higher magnitudes by the end of the century. However, it should be noted that autumn floods are generally only about half as high as spring floods in Sweden, except in the south, where the autumn and winter flows are normally larger. This also explains why this change in flow regime was not detected when focusing solely on annual maximum values for the whole country, which are dominated by the spring peak caused by snowmelt. There was a notable increase in the observed autumn peaks over the last 50 years, whereas the climate assessment with Hadley forcing revealed the largest increase in trend in the future. These results indicate an ongoing shift in flow regime, which can be attributed to flow-generating processes; by comparison, there will be less impact from floods generated by snowmelt in the spring and more frequent floods caused by intensive rainfall during the rest of the year.

3.6 Combining results to detect long-term changes in high flows

Assessing the past 100 years, we found no significant trends and only very small mean deviation in maximum daily flows (Table 2). The mean deviation for the autumn floods versus the reference period at the 69 river gauges was 9%, which means that the reference period was not representative of autumn floods, as can also be seen in Fig. 2. In contrast to the results for the last 50 years (Fig. 6), we found a negative trend in the autumn high flows for the last 100 years, although this was not statistically significant according to the trend test.

Using 100 years of climate model data with future projections revealed significant trends in upcoming changes. Both projections detected trends in the same direction but of different magnitudes and significance. The annual high flows showed a declining trend, which was even more pronounced when the analysis was limited to spring peaks. S-HYPE using Echam forcing indicated the largest negative trend entailing on average more than -2% reduction in spring peaks each decade. Conversely, there were positive trends in future autumn peaks, especially when using Hadley forcing, which resulted in a +3% increase per decade. Both these trends for the future were significant at $P = 0.05$, which confirmed the visual inspection of changes in flow regime in the Gauss curves (Fig. 7).

Figure 7 shows that there have been large long-term climate-induced oscillations in maximum daily flows during the last 100 years, which are expected to continue over the next hundred years. Furthermore, the oscillations in flood frequency were larger in past observations than in the future projections, although this might represent an artefact of the grid size used in the climate projections, which could have underestimated local extremes. Future long-term trends were consistent between climate projections, but each trend was statistically significant in only one projection (cf. Table 2).

Assessment of the seasonal cycle of high flow distribution in selected catchments indicated a temporal shift in maximum daily flows between the past, present, and future (Fig. 8). The last 25 years (present) have been warm and wet, and shown a tendency toward the results of the climate projections. Note that the diagrams in Fig. 8 represent time periods of different length and show absolute values instead of changes, and hence are not directly comparable but merely illustrate temporal changes during an average year. For the future, the results suggest that daily high flows occur about one month earlier during the spring in the north-central part of Sweden and become more frequent in the south, probably due to less snow accumulation in the south and at low altitudes.

Figure 8 also indicates that the spatial pattern of flow regimes across the country may change in some locations. There is a distinct border between snow-driven high flows in north-central Sweden and rain-driven high flows in the southern part of the country. For instance, the Fyrisån River is located in the north-central region where there has normally been a distinct snowmelt peak during spring in the past, but this is no longer apparent over the last 25 years or in the simulations of the future. Thus the climate impact on floods might be much more significant in some specific rivers compared to the average for the country. For a river such as the Fyrisån, this means that the risk of floods will be lower in the future than it was in the past.

4. Discussion

4.1 Changes in high flows in Sweden

This study revealed that pronounced shift in magnitudes of high flows induced by climate change has not yet been recorded nor is it expected in Sweden. However, our investigation focused on rivers, not on the small scale flooding caused by changes in intense local precipitation, which may have a greater impact in the future (e.g., in urban areas) (Arnbjerg-Nielsen et al., 2013; Olsson and Foster, 2014). We found a small, albeit not statistically significant, negative trend in river high flow indicating a 0.4% decrease in 10-yr flood frequency each decade. This confirms previous findings reported by Wilson et al. (2010) showing a decrease in peak flow events in long time series from Sweden, Finland, and parts of Denmark, but an increase in series from western Norway and Denmark. However, the changes we detected by using future climate projections in Sweden were statistically significant ($P = 0.05$) and in some cases of greater magnitude. It seems that annual daily maximum high flows may decrease by -1% per decade in the future, whereas autumn flows may increase by +3%, but the trends are far from linear. Assessing the maximum deviation versus the reference period shows that 1961-1990 cannot be used as a reference period in the future. Most design variables for infrastructure in Sweden are based on this period, and thus they must be recalculated using a new reference period to adapt them to climate change. Unfortunately, considering the past century, it also appears that 1961-1990 was not particularly representative of natural variability, especially regarding autumn floods.

Merging Gauss curves using both 100 years of observations and 100 years of climate projections clearly visualized the relative changes in and influence of long-term oscillations (Fig. 7). This combined analytical approach using both actual observations and model results simultaneously provides a broader understanding of natural versus accelerated changes in long time series. Applying shorter time scales to observed climate data gave a very different picture. For instance, when we used the 1960s as the starting point for a 50-year analysis

(Figs. 5 and 6), it seemed that the trend toward increased autumn floods was already very strong at that time, but this trend disappeared when we used 100 years of observations (cf. Table 2 and Fig. 7). This demonstrates that a period of 50 years is insufficient to detect trends in the Swedish climate. Lindström and Bergström (2004) found that trend detection is very sensitive to starting and ending years, which agrees with findings from other climate regions (e.g., Hannaford et al., 2013; Yue et al., 2012; Chen and Grasby, 2009).

In contrast to the trend analysis, our evaluation of annual dynamics and specific catchments revealed more radical changes in high flows. The earlier spring floods in the north-central part of Sweden, more frequent high flows in the south, and even disappearing spring peaks (Fig. 8) could be attributed to less snow accumulation in the south and at low altitudes. Similar findings have been made in Austria, where runoff trends could also be linked to altitude within catchments and attributed to changes in various processes dominating at different elevations (Kormann et al., 2014).

Spatial patterns can be noisy and make it difficult to detect overall trends due to local events. We used 69 gauging sites in our study and considered the mean of relative deviation (not absolute values) as representative of the country. Our frequency analysis also illustrated the spatial extent of specific high flow events. Originally, four hydroclimate regions were included in the evaluation (Fig. 1), but the results concerning observed changes differed very little between those regions, which were therefore considered to be too small to represent climate change. However, the projections of future climate differed markedly between the north and the southernmost regions (Figs. 4 and 8), for example, the positive trend in autumn flows with Hadley forcing was noted mainly in the north. Trend detection was based solely on spatially aggregated results for the entire domain, because we considered the discrepancies in the projections on a local or regional level to be too large to allow high-resolution analysis. Nevertheless, the observed high flow during the last 25 years did show a slight tendency toward the temporal changes that were suggested by the projections for individual catchments (Fig. 8). In addition, separation between rain fed and snow generated high flows could have been another basis for regional analysis. We therefore suggest a more thorough analysis for clustering catchments with similar behavior in future studies of regional changes within Sweden.

Results regarding the impact of climate change on frequency and intensity of floods in northern European countries are also available in the literature. Dankers and Feyen (2008) and Hirabayashi et al. (2008) indicated a decrease in water flow, whereas Lehner et al. (2006) suggested an increase, and Arheimer et al. (2012) projected very little overall change in water discharge to the Baltic Sea. Discrepancies between conclusions regarding the future can arise due to uncertainties in GCMs, downscaling methods, or hydrological models (e.g., Bosshard et al., 2013; Donnelly et al., 2014; Hall et al., 2014).

4.2 Methodological uncertainties

Both the use of observations from the past and modeling of the future involve uncertainties. Observed time series of river flow archived as part of Swedish national monitoring are calculated using measurements of water level and a traditional rating curve based on an observed relationship between water level and flow at each gauging site. Hence, each rating curve includes a number of variables that must be determined, and it is well known that rating curves may change over time (e.g., Tomkins, 2014; Westerberg, 2011) or can be overly simplistic due to hydraulic conditions at the gauging site (Le Coz et al., 2014). The monitored sections of the Swedish rivers at the gauges are considered to be rather stable, but a recent updating of a rating curve, after construction work at the gauging site, included changing the estimated water flow by approximately 30%. When rating curves are updated, the historical flow is also reconstructed. Nonetheless, this can be a major source of error in all analyses using observations of river flow. Extreme high flows are more uncertain than normal conditions, because in such cases the flow can be outside the calibrated range of the rating curve, and the flowing water may take new paths that bypass the gauging station. In Sweden, ice jam is another common monitoring problem, and hence observed time series are corrected for such blockage and reconstructed annually. These corrections may influence estimates of spring peaks in some of the northern rivers and represent another source of uncertainty.

It is even more difficult to monitor and model precipitation, because observations are influenced by changes in vegetation, wind, snow-/rainfall, and monitoring equipment. Furthermore, the monitoring technique employed at the beginning of last century probably underestimated precipitation (Lindström and Alexandersson, 2004). Experience has also shown that using the 4-km precipitation grid for operational hydrology, as done in our study, underestimates precipitation in the mountains of Sweden by some 10-20%. Accordingly, use of this grid as a source of observed climate data will obviously affect hydrological model results. Our validation of high flows in S-HYPE indicated median absolute errors of 15% and -3.5% underestimation in unregulated rivers (Fig. 3). Also, Bergstrand et al. (2014) have reported that after updating the S-HYPE model with gauged flow for national statistics and design variables, the mean high flows were underestimated by 5% at the 400 gauging stations, including those in regulated rivers. Clearly the underestimation of high flow is affected by the underestimated precipitation.

Major uncertainties associated with estimating future floods are related to the effects and interactions of the following components in the model chain (e.g., Bosshard et al., 2013): (1) climate model projections; (2) downscaling/bias correction techniques; (3) hydrological model uncertainties in the region studied.

4.2.1. Climate models

The discrepancy we found between our climate model projections indicates pronounced uncertainty of the local results, and trends in climate signals were often in opposite directions in the projections. It is well known that precipitation patterns in climate models differ considerably for different parts of Europe (e.g., van Ulden and van Oldenborgh, 2013), and this variability is further increased when extreme events are simulated by GCMs and RCMs (e.g., Blöschl et al., 2007). Hence the calculations performed are highly uncertain, and the findings concerning this part of the world should be approached with caution. Therefore, we limited our analysis to aggregated results concerning changes in floods in Sweden. It is normally recommended that decisions and impact modeling be based on the ensemble mean from many different climate projections (e.g., Bergström et al., 2012), but it is not known how much this will actually reduce the overall uncertainties. Ensemble runs correspond to a "sensitivity analysis" (inter-comparison of models) and not to uncertainty estimation in the statistical sense. Ensembles can also be biased by using many different versions of a particular model, and the GCMs/RCMs often include similar descriptions of the physics. In addition, some processes are not well represented in any climate model.

4.2.2. Downscaling and bias correction

Statistical downscaling and bias correction techniques involve empirical correction of simulated climate variables (e.g. precipitation and temperature) by fitting simulated means and quantiles to the available observations and applying the same correction to future simulations (e.g., Yang et al., 2010). Consequently, it is assumed that the observed biases in the mean and variability of those climate parameters are systematic and will be the same in the future, but it remains to be determined whether the climate model errors are static over time (Maraun et al., 2010). Use of bias correction methods leads to better fit of the hydrological model output, narrower variability bounds, and improved observed runoff regimes compared to uncorrected climate model data (Bosshard, 2011; Teutschbein and

Seibert, 2012). Nevertheless, bias correction can also introduce inconsistency between temperature and precipitation, which strongly affects simulation of snow variables (Dahné et al., 2013) and thereby also influences predictions of future floods. Furthermore, bias correction is very sensitive to the reference dataset applied, and thus conclusions regarding the hydrological impact of climate change may vary considerably even if all other aspects are kept constant (Donnelly et al, 2013). Therefore, Donnelly et al. (2014) have urged that, in addition to uncertainties in the climate model and scenario, uncertainties in the bias correction methodology and the impact model should be taken into account in studies concerning the impact of climate change.

4.2.3 Uncertainties in hydrological models of the studied region

Hydrological models are normally evaluated in relation to observed data, and uncertainties are well known and recognized. However, assessments of such models rarely focus on the skills required to predict climate change impact on a process level. The latest version of S-HYPE (2012) has an average Nash and Sutcliffe (1970) Efficiency (NSE) value of 0.81 for 200 stations unaffected by regulation and an average relative volume error of ± 5% for the period 1999-2008. For all 400 gauging stations, including both regulated and unregulated rivers, the average NSE is 0.70. All calibration criteria have some drawbacks, and one problem associated with NSE is that it focuses on timing and its use in optimization can thus underestimate the magnitude of high flow if the timing is not perfect.

The S-HYPE model is also assumed to be valid for ungauged basins, as has been confirmed by values from blind tests for independent gauging stations being comparable to those calibrated for groups of similar catchments (Arheimer and Lindström, 2013). S-HYPE captures hydroclimatic variability across Sweden, even though the gradients in temperature and precipitation in this country are larger than the estimated change in climate projections. However, variables that are sensitive to temperature (e.g., evapotranspiration) should be validated, in particular to ascertain whether their parameters are realistic for a changing climate. Use of several impact models is also recommended. For instance, Stahl et al. (2012) found that the mean of an ensemble of eight global hydrological models of Europe provided the best representation of trends in the observations.

The present scenarios consider changes in atmospheric emissions and concentrations of climate gases. However, in the future additional changes may well occur in other drivers of the hydrological regime, such as land use and vegetation, or construction work in river channels (Merz et al., 2012), which can also have a large impact on flood generation (e.g., Hall et al., 2014) and add uncertainties to predictions regarding flood frequency and magnitude. As described elsewhere (Arheimer and Lindström, 2014), we recently reconstructed the total impact of Swedish hydropower on the river water regime, which showed that spring peaks have decreased by 15% on a national scale. Hydropower in this country was established mainly from 1910 to 1970, and this anthropogenic alteration of the water resources has had a larger impact on river high flow than could be expected from climate change.

4.3 Gauss filtering

Statistical trend analyses were performed using discrete values of annual high flows, whereas visual inspections were conducted using a Gauss filter with a standard deviation corresponding to a moving average of 10 years. Gauss filtering dampens the effect of individual years and facilitates visual discrimination between the trends and oscillations. The filter does not remove all noise, and some oscillations also remain in a random dataset. However, a Gauss filter does not introduce any new oscillations. This is exemplified by the difference between periods in Fig. 2, which is real and not an artefact introduced by the filtering. For instance, the 1970s were dry in practically all of Sweden, whereas the 1920s, 1980s, and 1990s were mostly wet and had more frequent high autumn flows. The same periods stand out in other Nordic countries as well. The Gauss filter does not introduce any new trends, as the ones shown in Fig.5, since it only averages the signal over time. Hence, the filter is used merely to smooth the signal and compute decadal averages without the disadvantages of an ordinary running average. The Gauss is a low-pass filter that removes most of the interannual variation, and thus makes it easier to discern changes with a longer time scale (e.g., decades). Interestingly, it seems that the same pattern of more persistent periods of drier and wetter years that has occurred in the past (and is not introduced as an artefact by filtering) is preserved in the climate projections for the future. When making such projections, it is very important not to analyze specific years, because climate models do not yet offer such predictability but can only identify general trends and fluctuations that are not necessarily in phase with the observed climate. Therefore, rather than to present specific years from climate impact modeling, we chose to show only the general tendencies that are illustrated more clearly by Gauss filtering.

5. Conclusions

The present results indicate that there will be some shifts in flood-generating processes in Sweden in the future, and rain-generated floods will have a more marked effect. It is also plausible that there will be a greater climate impact on specific rivers than on the average for the entire country. Uncertainties and simultaneous changes from drivers other than climate must also be accounted for, although our findings do show the following:

- Changes in annual maximum daily flows in Sweden oscillate between clusters of years in relation to observed variability in weather, but no significant trend can be discerned over the past 100 years. We found a small tendency toward a decrease in high flows considering both magnitude and 10-yr flood frequency, but these results were not statistically significant.

- Temperature is the strongest driver of river high flows, because these events are related to snowmelt in most of Sweden. It is possible that the annual daily maximum flows will decrease in the future, mainly due to lower snowmelt peaks in spring as the result of earlier spring flood. In contrast, more intense rainfall and less snow accumulation may lead to increased autumn and winter flows.

- The temporal pattern of future daily high flows may shift in time and spring floods may occur approximately one month earlier in the north-central part of Sweden and more frequent high flows in the south due to less snow accumulation in the south and at low altitudes. Observations from the last 25 years have already shown a tendency toward this projected change.

- The spatial pattern across the country indicates a boundary between snow-driven high flows in north-central Sweden and rain-driven high flows in the south. This boundary may move to higher latitudes and altitudes with extension of the area with less common spring peaks and lower high flow.

6. Acknowledgements

This study was performed at the SMHI Hydrological Research Unit, where much work is done jointly to take advantages of previous studies and several projects conducted in parallel by the group. Hence input from individuals other than the authors was essential for the background material, and in particular we acknowledge the work performed by Johan Strömqvist and Thomas Bosshard for future climate change projections. Funding was provided by Swedish research councils: analysis of recorded flows was done with grants from HUVA/Elforsk, and assessments using model scenarios were funded by the projects Hydroimpacts2.0 (Formas) and CLEO (Swedish EPA). Our study will contribute to the IAHS Scientific Decade "Panta Rhei" initiative concerning changes in hydrology and society, and its working group on floods. S-HYPE results and gauged time-series, as well as tools for model uncertainty check and maps of climate scenario estimates, are available for inspection and free downloading at http://vattenwebb.smhi.se/. Finally, the authors thank Dr. Juraj Parajka, an anonymous reviewer, and the Editor Dr. Stacey Archfield for valuable comments on the manuscript.

7. References

Arheimer, B. and Lindström, G.: Implementing the EU Water Framework Directive in Sweden. Chapter 11.20 in: Bloeschl, G., Sivapalan, M., Wagener, T., Viglione, A. and Savenije, H. (Eds). Runoff Predictions in Ungauged Basins – Synthesis across Processes, Places and Scales. Cambridge University Press, Cambridge, UK. (p. 465) pp. 353-359. 2013.

Arheimer, B. and Lindström, G.: Electricity vs Ecosystems – understanding and predicting hydropower impact on Swedish river flow. Evolving Water Resources Systems: Understanding, Predicting and Managing Water–Society Interactions. Proceedings of ICWRS2014, Bologna, Italy, June 2014. IAHS Publ. No. 364. 2014.

Arheimer, B., Dahné J., and Donnelly, C.: Climate change impact on riverine nutrient load and land-based remedial measures of the Baltic Sea Action Plan. Ambio 41 (6):600-612. 2012.

Arnbjerg-Nielsen, K., Willems, P., Olsson, J., Beecham, S., Pathirana, A., Bülow Gregersen, I., Madsen, H., and V.T.V. Nguyen. Impacts of climate change on rainfall extremes and urban drainage systems: a review. Water Sci. Technol., 68:16-28, doi:10.2166/wst.2013.251. 2013.

Bates, B.C., Kundzewicz, Z.W., Wu, S., Palutikof, J. (Eds): Climate Change and Water. Technical Paper of the Intergovernmental Panel on Climate Change. IPCC Secretariat, Geneva, 210 pp. 2008.

Bergstrand, M., Asp, S., and Lindström, G.: Nation-wide hydrological statistics for Sweden with high resolution using the hydrological model S-HYPE. Hydrology Research (in press). 2014

Bergström, S., Andréasson, J. and Graham L.P.: Climate adaptation of the Swedish guidelines for design floods for dams. ICold 24th Congress Kyoto 2012, Q94. 2012.

Blöschl, G., Ardoin-Bardin, S., Bonell, M., Dorninger, M., Goodrich, D., Gutknecht, D., Matamoros, D., Merz, B., Shand, P., and Szolgay, J.: At what scales do climate variability and land cover change impact on flooding and low flows?, Hydrol. Process., 21, 1241–1247, doi:10.1002/hyp.6669. 2007.

Bosshard, T., Carambia, M. Goergen, K. Kotlarski, S. Krahe, P. Zappa, M. and Schar, C.: Quantifying uncertainty sources in an ensemble of hydrological climate-impact projections, Water Resour. Res., 49, 1523-1536, doi: 10.1029/2011WR011533. 2013.

Bosshard, T., Kotlarski, S., Ewen, T., and Schär, C.: Spectral representation of the annual cycle in the climate change signal, Hydrol. Earth Syst. Sci., 15, 2777-2788, doi:10.5194/hess-15-2777-2011. 2011.

Bosshard, T., Olsson, J.: Comparison of the two climate projections in CLEO to a larger ensemble, CLEO report , 20 pp. http://www.cleoresearch.se/publications/cleoreports 2014.

Chen, Z. and Grasby, S. E.: Impact of decadal and century-scale oscillations on hydroclimate trend analyses, J. Hydrol., 365, 122–133, doi:10.1016/j.jhydrol.2008.11.031. 2009.

Collins, M., Booth, B.B.B., Harris, G.R., Murphy, J.M., Sexton, D.M.H., Webb, M.J.: Towards quantifying uncertainty in transient climate change, Climate Dynamics, 27, 127-147, DOI 10.1007/s00382-006-0121-0. 2006.

Dahné, J., Donnelly, C, and Olsson, J.: Post-processing of climate projections for hydrological impact studies: how well is the reference state preserved? IAHS Publ. 359. 2013.

Dankers, R. & Feyen, L.: Climate change impact on flood hazard in Europe: An assessment based on high-resolution climate simulations. Journal of Geophysical Research: Atmospheres, 113, D19105. doi:10.1029/2007JD009719. 2008.

Donnelly, C., Arheimer, B., Bosshard, T. and Pechlivanidis, I.: Uncertainties beyond ensembles and parameters – experiences of impact assessments using the HYPE model at various scales. Proceedings ImpactsWorld 2013. International Conference on Climate Change Effects, Potsdam, May, 2013

Donnelly, C., Yang, W. and Dahné, J.: River discharge to the Baltic Sea in a future climate. Climatic Change 122(1-2): 157-170, doi: 10.1007/s10584-013-0941-y, 2014.

Hall, J., Arheimer, B., Borga, M, Brázdil, R., Claps, P., Kiss, A., Kjeldsen, T.R., Kriaučiūnienė, J., Kundzewicz, Z., Lang, M., Llasat, M.C., Macdonald, N., McIntyre, N., Mediero, L., Merz, B., Merz, R., Molnar, P., Montanari, A., Neuhold, C., Parajka, J., Perdigão, R.A. P., Plavcová, L., Rogger, M., Salinas, J.L., Sauquet, E., Schär, C., Szolgay, J., Viglione, A., and Blöschl, G. Understanding Flood Regime Changes in Europe: A state of the art assessment, Hydrol. Earth Syst. Sci. (HESS), 18, 2735–2772 (doi:10.5194/hess-18-2735-2014). 2014.

Hannah, D. M., Demuth, S., Lanen van, H. A. J., Looser, U., Prudhomme, C., Rees, G., Stahl, K., and Tallaksen, L. M.: Large-scale river flow archives: importance, current status and future needs. Hydrol. Processes 25(7): 1191–1200, doi:10.1002/hyp.7794. 2010.

Hannaford, J., Buys, G., Stahl, K., and Tallaksen, L. M. The Influence of decadal-scale variability on trends in long European streamflow records. Hydrology and Earth Systems Science 17:2717-2733, doi:10.5194/hess-17-2717-2013. 2013.

Hellström, S. and Lindström, G.: Regional analys av klimat, vattentillgång och höga flöden. SMHI Report Hydrologi No 110 (in Swedish) 2008.

Hirabayashi, Y., Kanae, S., Emori, S., Oki, T., and Kimoto, M.: Global projections of changing risks of floods and droughts in a changing climate, Hydrolog. Sci. J., 53, 754–772, doi:10.1623/hysj.53.4.754. 2008

Huntington, T. G.: Evidence for intensification of the global water cycle: review and synthesis, J. Hydrol., 319, 83–95, doi:10.1016/j.jhydrol.2005.07.003. 2006

Johansson, B.: Estimation of areal precipitation for hydrological modelling in Sweden. PhD Thesis, Earth Sciences Centre, Dept Phys. Geog., Göteborg University. Sweden. 2002.

Johns, T.C., Gregory, J.M, Ingram, W.J, Johnson, C.E., Jones, A., Lowe, J.A., Mitchell, J.F.B., Roberts, D.L., Sexton, D.M.H, Stevenson, D.S., Tett, S.F.B., Woodage, M.J.: Anthropogenic climate change for 1860 to 2100 simulated with the HadCM3 model under updated emissions scenarios, Climate Dynamics, 20, 583–612, DOI 10.1007/s00382-002-0296-y. 2003.

Kjellström, E., Nikulin, G., Hansson, U., Strandberg, G., Ullerstig, A.: 21st century changes in the European climate: uncertainties derived from an ensemble of regional climate model simulations. Tellus 63A:24-40. doi. 10.1111/j.1600-0870.2010.00475.x. 2011.

Kormann, C., Francke, T., Renner, M. and Bronstert, A.: Hydrol. Earth Syst. Sci. Discuss., 11, 6881–6922. www.hydrol-earth-syst-sci-iscuss.net/11/6881/2014/doi:10.5194/hessd-11-6881-2014. 2014.

Kundzewicz, Z.W. et al.: Freshwater resources and their management. In: Parry M.L., et al., eds., Climate Change 2007: Impacts, Adaptation and Vulnerability. Contribution of Working Group II to the Fourth Assessment Report of the Intergovernmental Panel on Climate Change, Cambridge, UK: Cambridge University Press. 2007.

Le Coz, J., Renard, B., Bonnifait, L., Branger, F. Le Boursicaud, R.: Combining hydraulic knowledge and uncertain gaugings in the estimation of hydrometric rating curves: A Bayesian approach. Journal of Hydrology 509(0): 573-587. 2014.

Lehner, B., Döll, P., Alcamo, J., Henrichs, T., and Kaspar, F.: Estimating the impact of global change on flood and drought risks in Europe: a continental, integrated analysis, Climatic Change, 75, 273–299, doi:10.1007/s10584-006-6338-4. 2006.

Lindström, G. and Alexandersson, H.: Recent mild and wet years in relation to long observation records and climate change in Sweden. Ambio, Volume 33(4-5):183-186. 2004.

Lindström, G. and Bergström, S.: Runoff trends in Sweden 1807-2002. Hydrological Sciences Journal, 49(1): 69-83. 2004.

Lindström, G., Pers, C.P., Rosberg, R., Strömqvist, J., Arheimer, B.: Development and test of the HYPE (Hydrological Predictions for the Environment) model – A water quality model for different spatial scales. Hydrology Research 41.3-4:295-319. 2010.

Maraun, D., et al.: Precipitation downscaling under climate change: Recent developments to bridge the gap between dynamical models and the end user, Rev. Geophys., 48, RG3003, doi:10.1029/2009RG000314. 2010.

Markonis, Y., Koutsoyiannis, D.: Climatic Variability Over Time Scales Spanning Nine Orders of Magnitude: Connecting Milankovitch Cycles with Hurst–Kolmogorov Dynamics, Survey in Geophysics, Surv Geophys, DOI 10.1007/s10712-012-9208-9. 2012.

Merz, B., Vorogushyn, S., Uhlemann, S., Delgado, J., and Hundecha, Y.: HESS Opinions "More efforts and scientific rigour are needed to attribute trends in flood time series", Hydrol. Earth Syst. Sci., 16, 1379–1387, doi:10.5194/hess-16-1379-2012. 2012

Montanari, A.: Hydrology of the Po River: looking for changing patterns in river discharge, Hydrol. Earth Syst. Sci., 16, 3739–3747, doi:10.5194/hess-16-3739-2012. 2012.

Murphy, J. M., Booth, B., Collins, M., Harris, G., Sexton, D., and Webb, M.: A methodology for probabilistic predictions of regional climate change from perturbed physics ensembles, Philos. T. Roy. Soc. A, 365, 1993–2028, doi:10.1098/rsta.2007.2077. 2007.

Nash, J. E. & Sutcliffe, J. V.: River flow forecasting through conceptual models. Part I. A discussion of principles. J. Hydrol. 10, 282–290. 1970.

Nakićenović, N., Alcamo, J., Davis, G., de Vries, B., Fenhann, J., Gaffin, S., Gregory, K., Grübler, A., et al.: Emission scenarios. A Special Report of Working Group III of the Intergovernmental Panel on Climate Change. Cambridge University Press, 599 pp. 2000.

Olsson, J., and Foster, K. Short-term precipitation extremes in regional climate simulations for Sweden: historical and future changes. Hydrol. Res., 45:479-489, doi:10.2166/nh.2013.206. 2014.

Parajka, J., Kohnová, S., Merz, R., Szolgay, J., Hlavová, K., and Blöschl, G.: Comparative Analysis of the seasonality of hydrological characteristics in Slovakia and Austria, Hydrolog. Sci. J., 54, 456–473, doi: 10.1623/hysj.54.3.456. 2009.

Petrow, T. and Merz, B.: Trends in flood magnitude, frequency and seasonality in Germany in the period 1951-2002, Journal of Hydrology, 371, 129-141. 2009.

Roeckner, E., R. Brokopf, M. Esch, M. Giorgetta, S. Hagemann, L. Kornblueh, E. Manzini, U. Schlese, and U. Schulzweida: Sensitivity of simulated climate to horizontal and vertical resolution in the ECHAM5 atmosphere model, J. Climate, 19, 3771-3791, 2006

Samuelsson, P., Jones, C.G., Willén, U., Ullerstig, A., Gollvik, S., Hansson, U., Jansson, C., Kjellström, E., Nikulin, G., Wyser, K.: The Rossby Centre Regional Climate model RCA3: Model description and performance. Tellus A 63:4-23. doi. 10.1111/j.1600-0870.2010.00478.x. 2011.

Schmocker-Fackel, P. and Naef, F.: More frequent flooding? Changes in flood frequency in Switzerland since 1850. Journal of Hydrology, 381(1-2), 1–8. 2010.

Stahl, K., Hisdal, H., Hannaford, J., Tallaksen, L. M., van Lanen, H. A. J., Sauquet, E., Demuth, S., et al.: Streamflow trends in Europe: evidence form a dataset of near-natural catchments. Hydrology and Earth System Sciences, 14, 2367–2382. 2010.

Stahl, K., Tallaksen, L. M., Hannaford, J. and van Lanen, H. A. J.: Filling the white space on maps of European runoff trends: estimates from a multi-model ensemble. Hydrological and Earth System Sciences, 16, 2035–2047. 2012.

Strömqvist, J., Arheimer, B., Dahné, J., Donnelly, C. and Lindström, G.: Water and nutrient predictions in ungauged basins – Set-up and evaluation of a model at the national scale. Hydrological Sciences Journal 57(2):229-247. 2012.

Teutschbein, C. and Seibert, J.: Bias correction of regional climate model simulations for hydrological climate-change impact studies: review and evaluation of different methods, J. Hydrol., 456, 12–29, doi:10.1016/j.jhydrol.2012.05.052. 2012.

Tomkins, K. M.: Uncertainty in streamflow rating curves: methods, controls and consequences. Hydrol. Process., 28: 464–481. doi: 10.1002/hyp.9567. 2014.

van Ulden, A.P. and van Oldenborgh, G.J.: Large-scale atmospheric circulation biases and changes in global climate model simulations and their importance for climate change in Central Europe Atm. Chem. Phys., 2006, 6, 863-881, sref:1680-7324/acp/2006-6-863. 2013.

Westerberg, I., Guerrero, J.-L., Seibert, J., Beven, K. J. and Halldin, S.: Stage-discharge uncertainty derived with a non-stationary rating curve in the Choluteca River, Honduras. Hydrol. Process., 25: 603–613. doi: 10.1002/hyp.7848. 2011.

Yang, W., Andréasson, J., Graham, L. P., Olsson, J., Rosberg, J., and Wetterhall, F.: Distribution based scaling to improve usability of regional climate model projections for hydrological climate change impacts studies, Hydrol. Res., 41, 211–229, doi:10.2166/nh.2010.004. 2010.

Yevjevich, V.: Probability and Statistics in Hydrology. Water Resources Publications, Fort Collins, Colorado, USA. 1972.

Yue, S., Kundzewicz, Z. W., and Wang, L.: Detection of changes, in: Changes in Flood Risk in Europe, edited by: Kundzewicz, Z. W., IAHS Press, Wallingford, UK, 387–434. 2012.

Tables

Table 1. Deviation (%) in relation to the mean for the reference period (1961-1990) and trends (slope in percent per decade) for annual anomalies in high flows at the 69 river gauges, using observed discharge from gauges and S-HYPE modeled discharge, the latter with Hadley or Echam forcing from observed climate and climate projections. Bold numbers indicate a significance level of $P = 0.05$ (Yevjevich, 1972).

	Full overlapping period 1961-2010		Independent period 1961-1980		Reference period 1961-1990		S-HYPE calibration period 1999-2008		DBS calibration period 1981-2010	
	Deviation mean (%)	Trend slope	Deviation mean (%)	Trend slope	Deviation mean (%)	Trend slope	Deviation mean (%)	Trend slope	Deviation mean (%)	Trend slope
69 gauge stations	1.2	0.0	1.9	0.7	0.2	0.1	-1.8	0.2	0.7	0.1
S-HYPE with:										
Obs climate	0.6	0.2	-2.1	1.1	0.0	0.5	-0.7	0.5	0.7	-0.2
Hadley climate	0.6	0.2	-2.9	**1.4**	0.0	**0.7**	6.5	-1.0	3.0	-0.2
Echam climate	-1.0	-0.1	1.2	-0.1	0.0	-0.1	-2.4	0.6	-2.5	0.2

Table 2. Summary of analysis of daily high flows in observed time series representing 100 years in the past and modeled time series for 100 years in the future. Deviation (%) in relation to the mean of the reference period (1961-1990) and trends (slope as percent per decade) are given for annual high flows, frequency of 10-yr flood, and spring and autumn flood. Bold numbers indicate a significance level of $P = 0.05$ (Yevjevich, 1972).

		Frequency of 10-yr flood		Annual high flow		High flows March-June		High flows July-Febr	
Data source	100 yrs Period	Fraction mean (No)	Trend slope	Deviation mean (%)	Trend slope	Deviation mean (%)	Trend slope	Deviation mean (%)	Trend slope
Observations in:									
69 gauge stations	1910-2010	12	-0.4	0.0	-0.3	3.0	0.0	8.9	-1.1
S-HYPE with:									
Hadley climate	2000-2100	12	0.4	-1.3	-0.4	-7.7	-1.1	19.9	**3.0**
Echam climate	2000-2100	8	-0.2	-8.5	**-1.3**	-15.3	**-2.1**	2.7	1.1

Figures

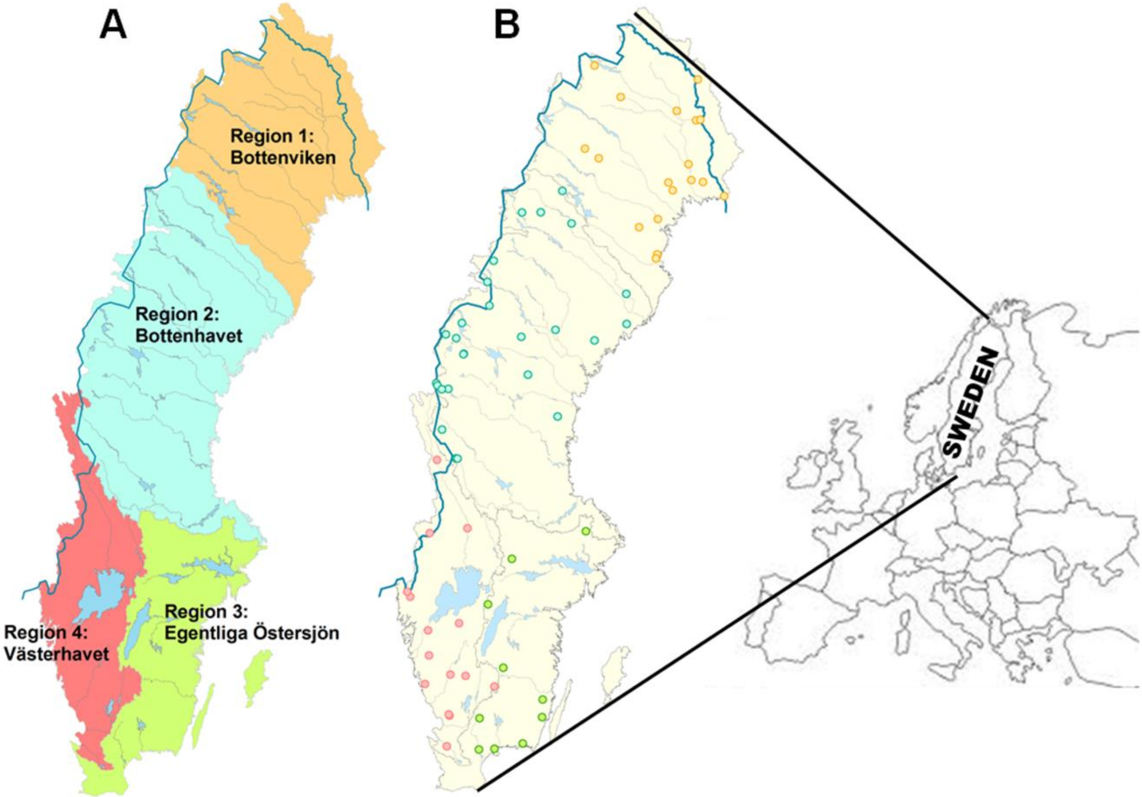

Figure 1. Maps showing (A) the four climate regions in Sweden and (B) locations of the 69 gauges with long-term records from unregulated rivers.

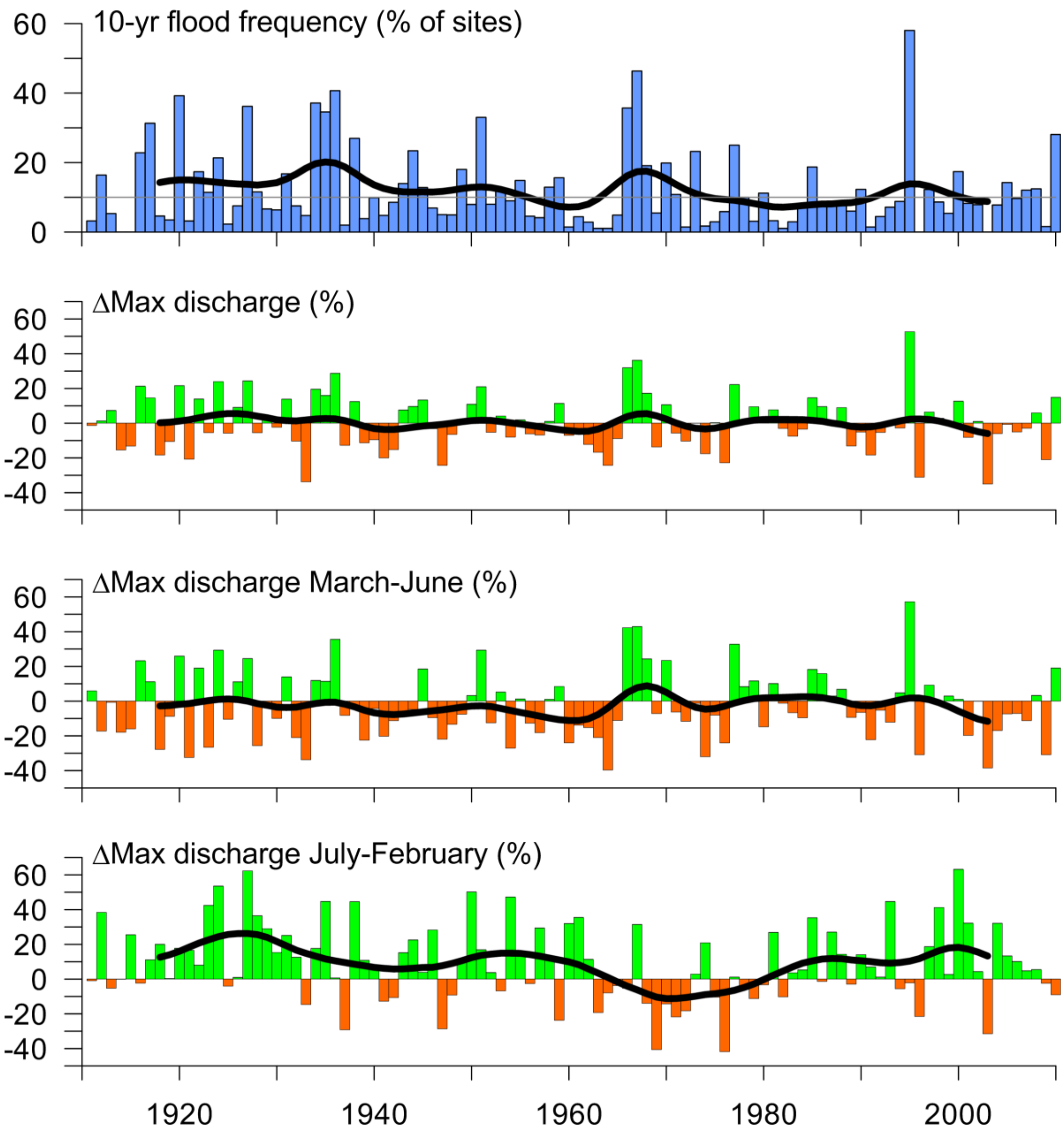

Figure 2. Observed annual high flow (1911-2010) *versus* reference period (1961-1990) for the 69 rivers, showing fractions of stations exceeding the 10-year flood each year, mean deviation in the magnitude of annual maximum daily discharge, and mean deviation in the magnitude of maximum daily discharge during March-June and July-February. The black line represents a 10-year Gauss filter.

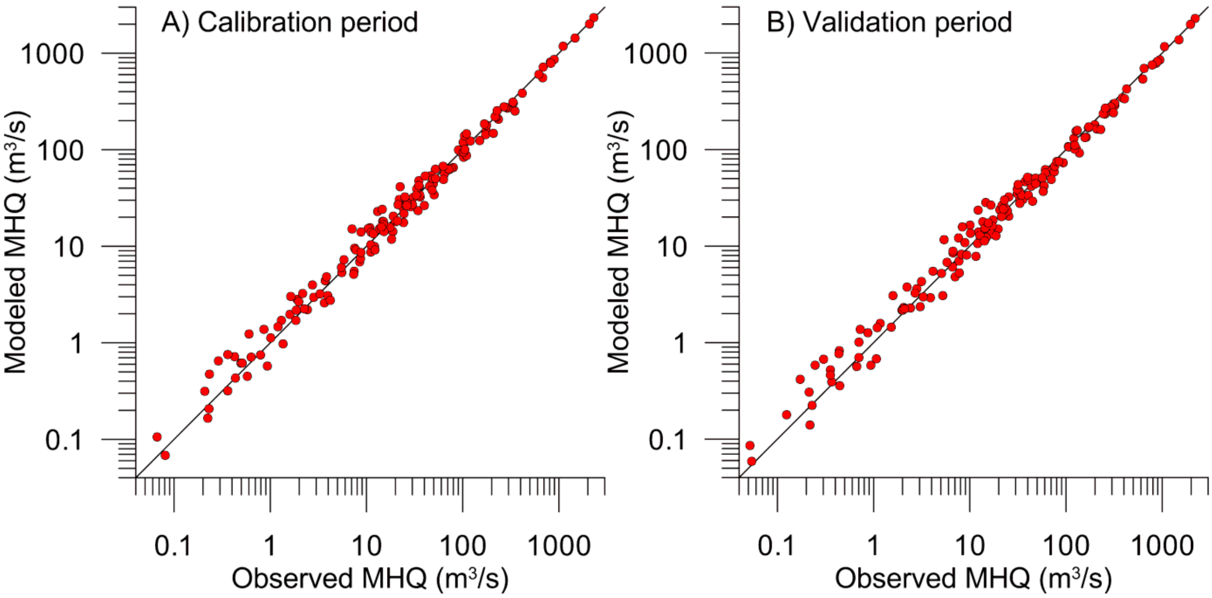

Figure 3. Observed versus predicted annual high flow from the S-HYPE model for (A) the calibration period (1999-2008) and (B) the validation period (1988-1998). MHQ = mean high flow.

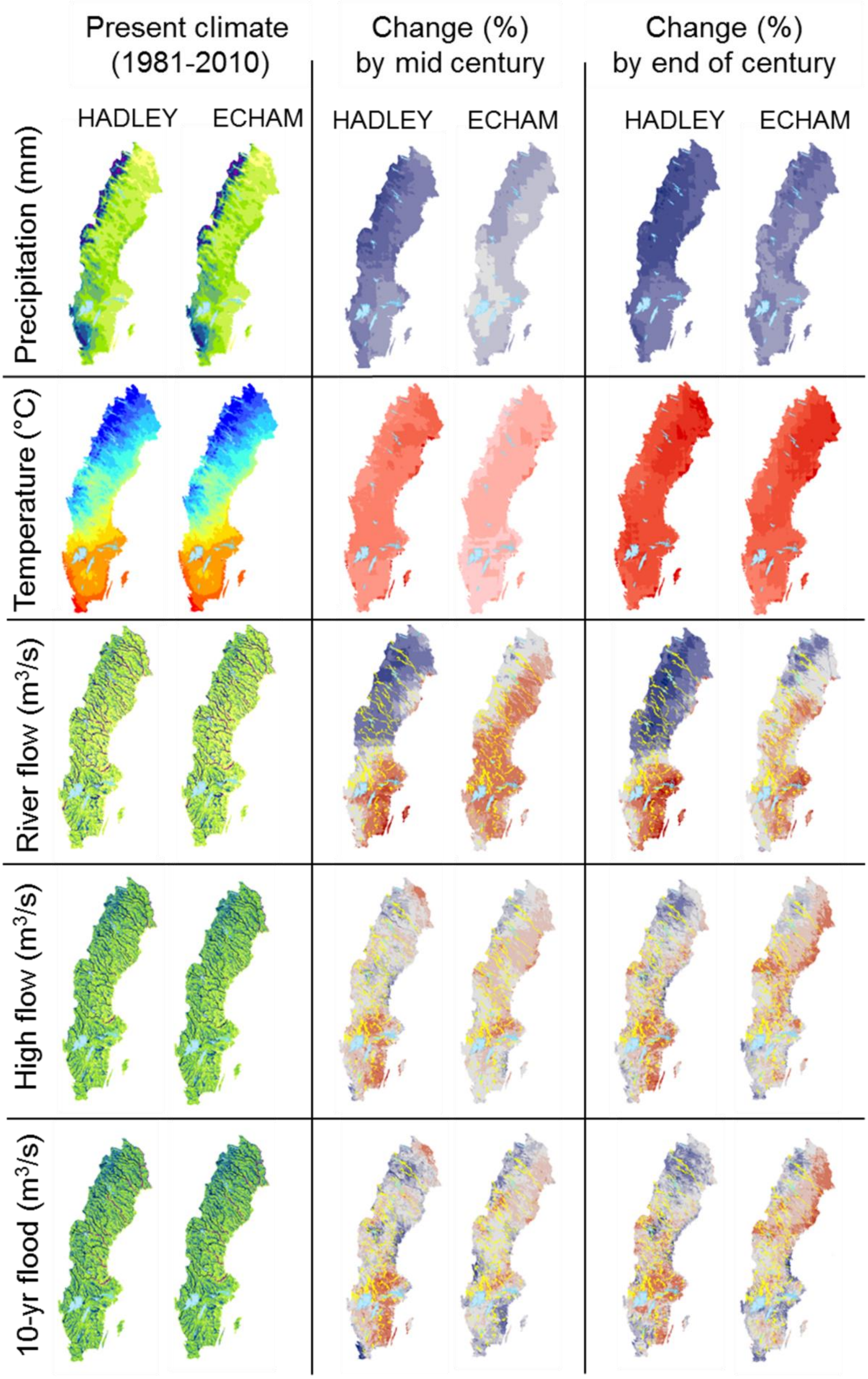

Figure 4. Spatial patterns of climate change impact across Sweden obtained using two downscaled and bias-corrected climate projections in S-HYPE. Mean values for mid century (2035-2065) and end of century (2071-2100) are compared with the mean for a reference period (1981-2010). Red indicates warmer/drier, and blue represents colder/wetter. Results are not shown for highly regulated rivers (yellow).

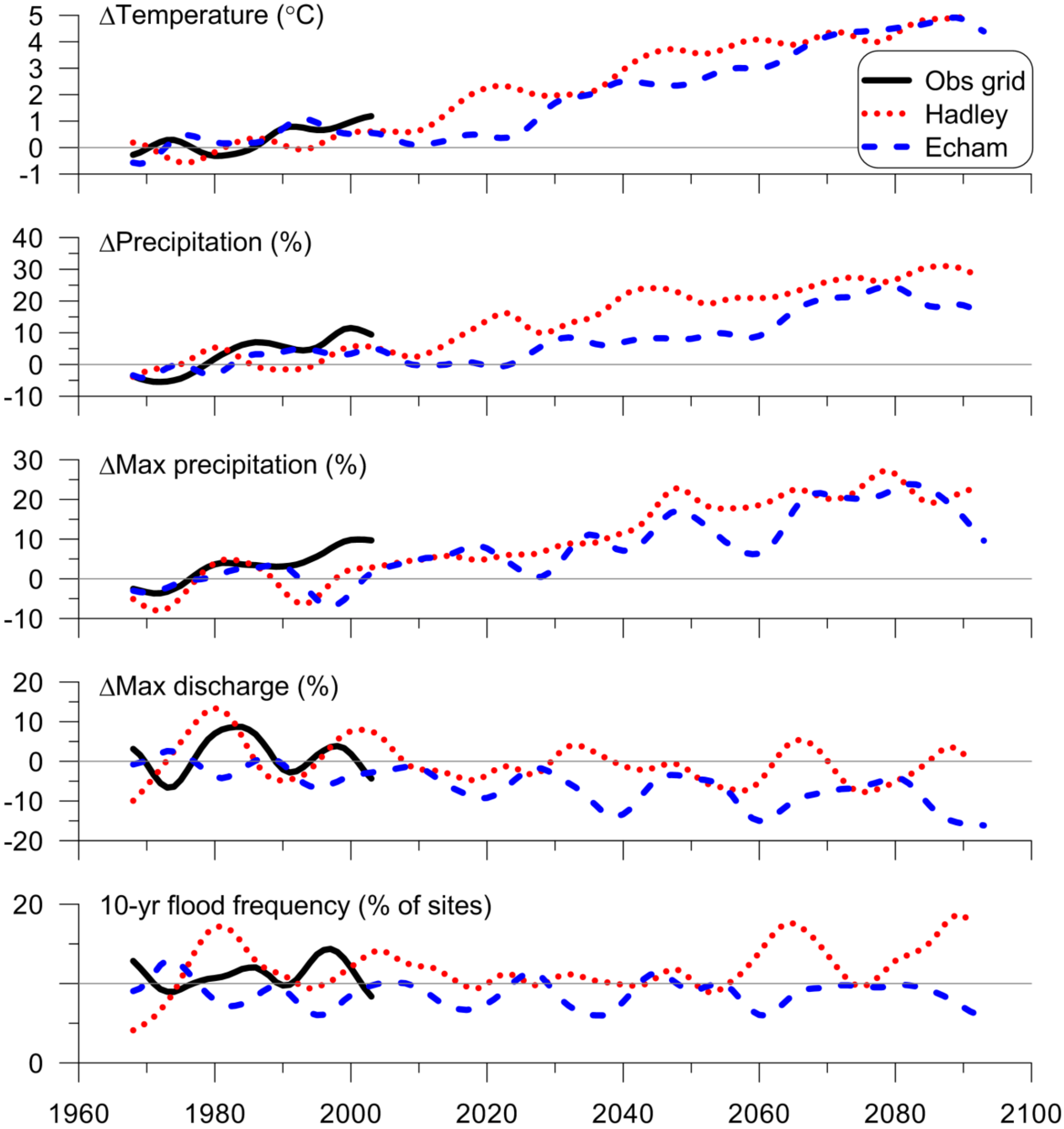

Figure 5. Modeled deviation (%) in annual regional estimates 1961-2100 *versus* the reference period (1961-1990) using S-HYPE for annual mean temperature and precipitation, maximum daily precipitation, annual daily high flow, and number of gauges exceeding the 10-year flood. A 10-year Gauss filter was used to filter annual results. Modeling was done with forcing data based on observations (Obs grid; solid lines) or on climate models (Hadley and Echam; dotted and dashed lines).

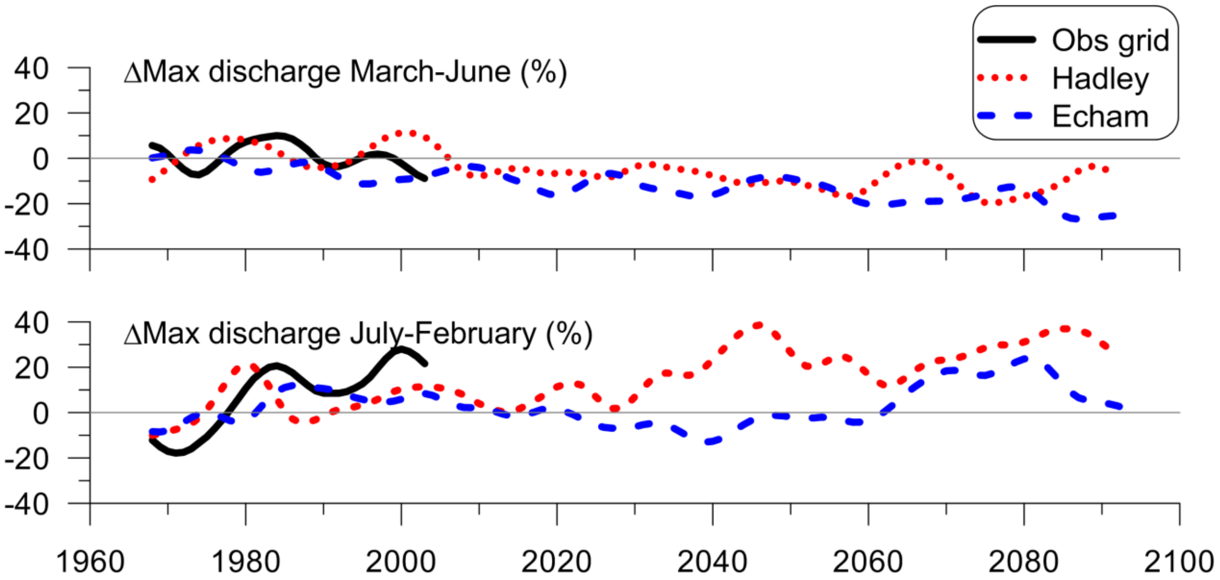

Figure 6. Modeled annual maximum daily flows during spring (top) and autumn/winter (bottom) for the period 1961-2100. Deviation (%) in magnitude *vs* reference period (1961-1990). The lines represent a 10-year Gauss filter for S-HYPE modeling using a forcing grid based on observations (Obs grid) or climate projections (Hadley and Echam).

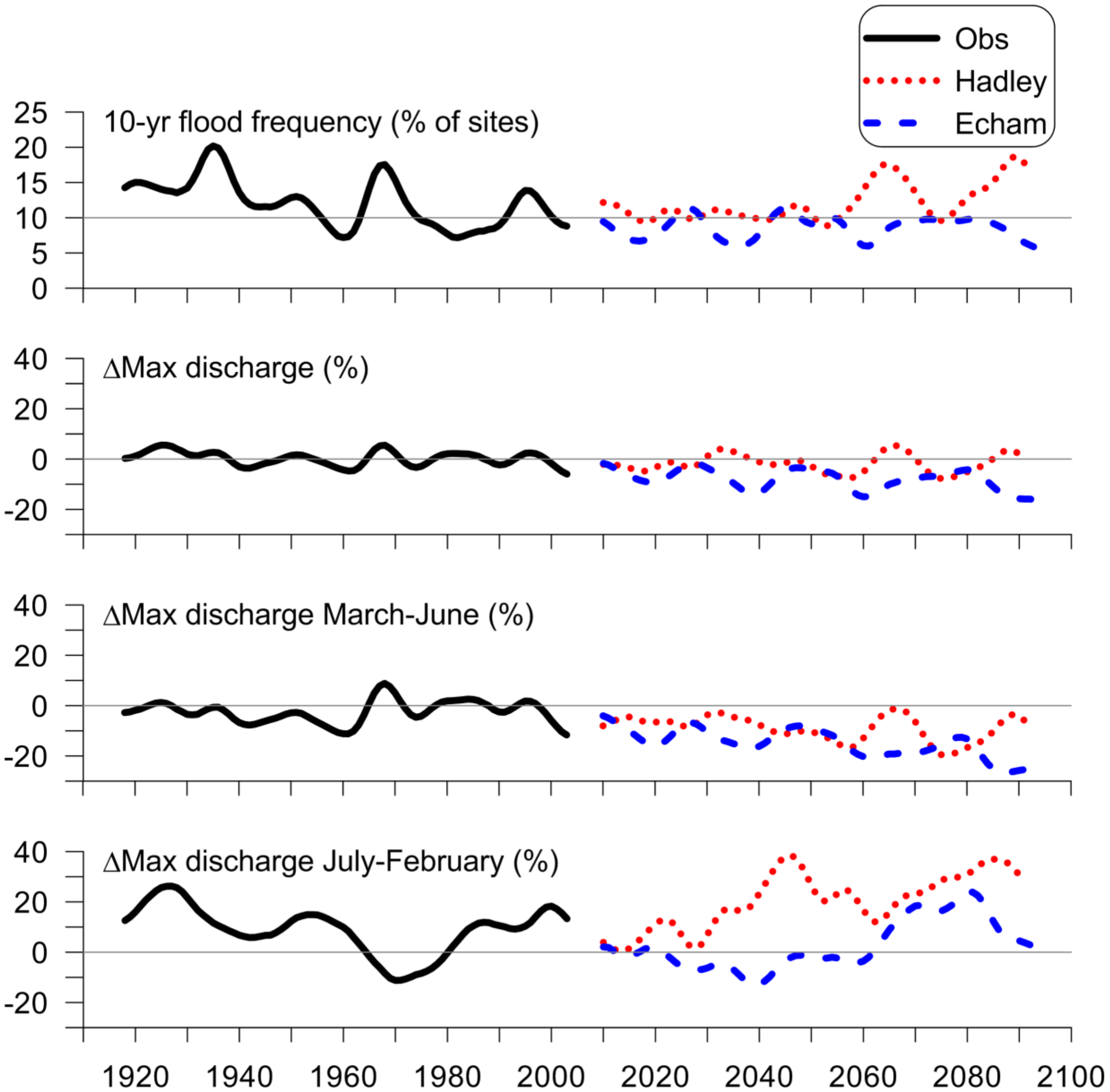

Figure 7. Merged time series of deviations (%) versus the mean for the reference period shown for actual observations (1910-2010) and modeling results (2010-2100) for past and future annual maximum high flows in Sweden. The lines represent a 10-year Gauss filter for observations (obs) and S-HYPE forced with climate projections (Hadley and Echam).

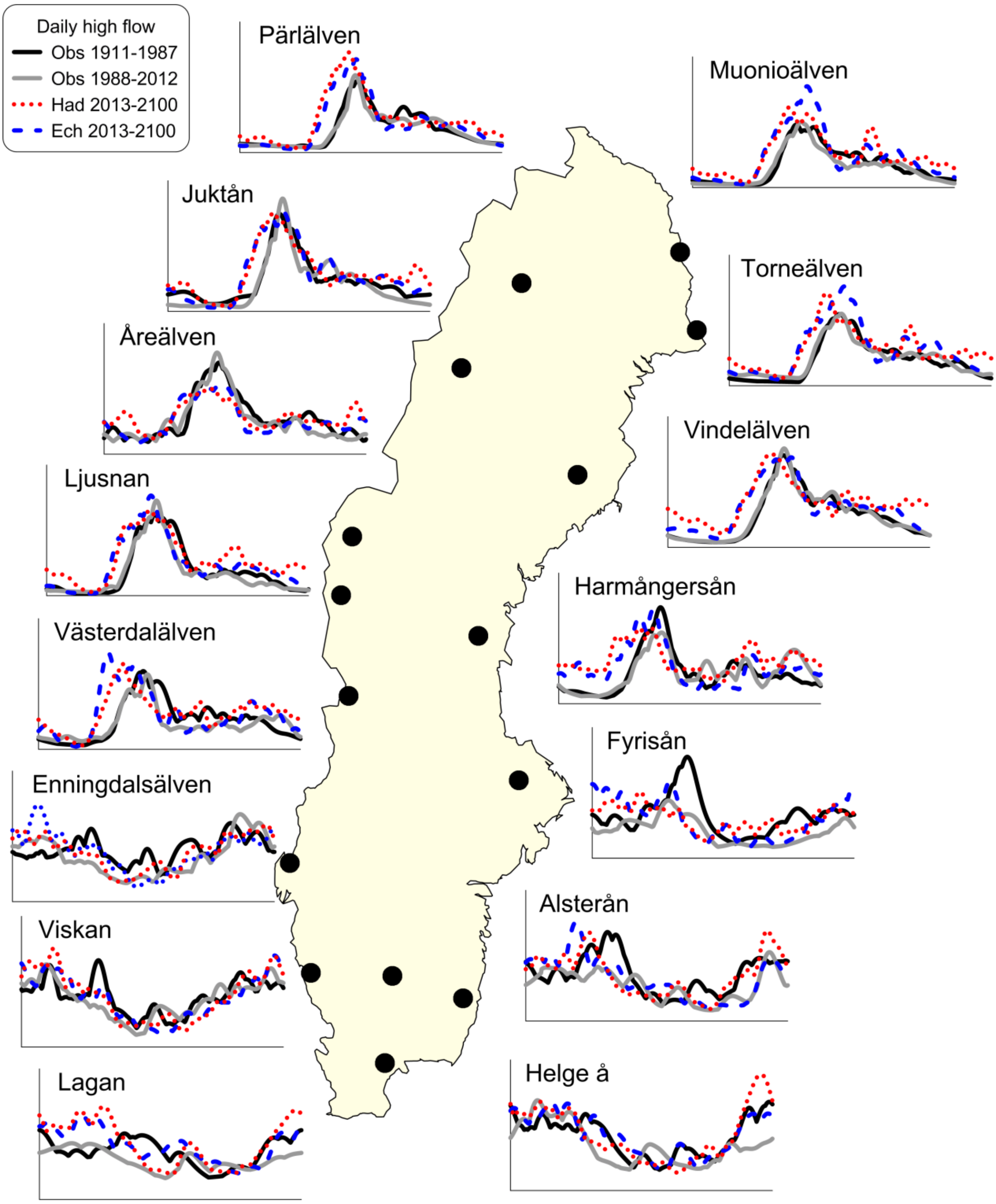

Figure 8. Annual distribution of daily high flow (Jan-Dec) in selected catchments across Sweden obtained using a 1-month Gauss filter for observed and projected time series. Note: magnitudes of observed and projected values are not comparable, only timing of high flows should be compared. Solid lines represent observations (Obs) for different time periods. Dotted and dashed lines represent S-HYPE modeling with forcing data from the climate models Hadley (Had) and Echam (Ech), respectively.