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Tyndall[°]Centre for Climate Change Research

Climate: Observations, projections and impacts: Germany

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The country reports were written by a range of climate researchers, chosen for their subject expertise, who were drawn from institutes across the UK. Authors from the Met Office and the University of Nottingham collated the contributions in to a coherent narrative which was then reviewed. The authors and contributors of the reports are as above. Developed at the request of:



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Climate: Observations, projections and impacts





We have reached a critical year in our response to climate change. The decisions that we made in Cancún put the UNFCCC process back on track, saw us agree to limit temperature rise to 2 °C and set us in the right direction for reaching a climate change deal to achieve this. However, we still have considerable work to do and I believe that key economies and major emitters have a leadership role in ensuring a successful outcome in Durban and beyond.

To help us articulate a meaningful response to climate change, I believe that it is important to have a robust scientific assessment of the likely impacts on individual countries across the globe. This report demonstrates that the risks of a changing climate are wide-ranging and that no country will be left untouched by climate change.

I thank the UK's Met Office Hadley Centre for their hard work in putting together such a comprehensive piece of work. I also thank the scientists and officials from the countries included in this project for their interest and valuable advice in putting it together. I hope this report will inform this key debate on one of the greatest threats to humanity.

The Rt Hon. Chris Huhne MP, Secretary of State for Energy and Climate Change



There is already strong scientific evidence that the climate has changed and will continue to change in future in response to human activities. Across the world, this is already being felt as changes to the local weather that people experience every day.

Our ability to provide useful information to help everyone understand how their environment has changed, and plan for future, is improving all the time. But there is still a long way to go. These reports – led by the Met Office Hadley Centre in collaboration with many institutes and scientists around the world – aim to provide useful, up to date and impartial information, based on the best climate science now available. This new scientific material will also contribute to the next assessment from the Intergovernmental Panel on Climate Change.

However, we must also remember that while we can provide a lot of useful information, a great many uncertainties remain. That's why I have put in place a long-term strategy at the Met Office to work ever more closely with scientists across the world. Together, we'll look for ways to combine more and better observations of the real world with improved computer models of the weather and climate; which, over time, will lead to even more detailed and confident advice being issued.

Julia Slingo, Met Office Chief Scientist

Introduction

Understanding the potential impacts of climate change is essential for informing both adaptation strategies and actions to avoid dangerous levels of climate change. A range of valuable national studies have been carried out and published, and the Intergovernmental Panel on Climate Change (IPCC) has collated and reported impacts at the global and regional scales. But assessing the impacts is scientifically challenging and has, until now, been fragmented. To date, only a limited amount of information about past climate change and its future impacts has been available at national level, while approaches to the science itself have varied between countries.

In April 2011, the Met Office Hadley Centre was asked by the United Kingdom's Secretary of State for Energy and Climate Change to compile scientifically robust and impartial information on the physical impacts of climate change for more than 20 countries. This was done using a consistent set of scenarios and as a pilot to a more comprehensive study of climate impacts. A report on the observations, projections and impacts of climate change has been prepared for each country. These provide up to date science on how the climate has already changed and the potential consequences of future changes. These reports complement those published by the IPCC as well as the more detailed climate change and impact studies published nationally.

Each report contains:

- A description of key features of national weather and climate, including an analysis of new data on extreme events.
- An assessment of the extent to which increases in greenhouse gases and aerosols in the atmosphere have altered the probability of particular seasonal temperatures compared to pre-industrial times, using a technique called 'fraction of attributable risk.'
- A prediction of future climate conditions, based on the climate model projections used in the Fourth Assessment Report from the IPCC.
- The potential impacts of climate change, based on results from the UK's Avoiding Dangerous Climate Change programme (AVOID) and supporting literature. For details visit: http://www.avoid.uk.net

The assessment of impacts at the national level, both for the AVOID programme results and the cited supporting literature, were mostly based on global studies. This was to ensure consistency, whilst recognising that this might not always provide enough focus on impacts of most relevance to a particular country. Although time available for the project was short, generally all the material available to the researchers in the project was used, unless there were good scientific reasons for not doing so. For example, some impacts areas were omitted, such as many of those associated with human health. In this case, these impacts are strongly dependent on local factors and do not easily lend themselves to the globally consistent framework used. No attempt was made to include the effect of future adaptation actions in the assessment of potential impacts. Typically, some, but not all, of the impacts are avoided by limiting global average warming to no more than 2 °C.

The Met Office Hadley Centre gratefully acknowledges the input that organisations and individuals from these countries have contributed to this study. Many nations contributed references to the literature analysis component of the project and helped to review earlier versions of these reports.

We welcome feedback and expect these reports to evolve over time. For the latest version of this report, details of how to reference it, and to provide feedback to the project team, please see the website at www.metoffice.gov.uk/climate-change/policy-relevant/obs-projections-impacts

In the longer term, we would welcome the opportunity to explore with other countries and organisations options for taking forward assessments of national level climate change impacts through international cooperation.

Summary

Climate observations

- Over the period 1960 to 2010 there was a consistent warming over Germany.
- There has been a substantial decrease in the frequency of cool nights and cool days since 1960 and a strong increase in the numbers of warm nights and warm days.
- There has been a general increase in summer temperatures averaged over the country as a result of human influence on climate, making the occurrence of warm summer temperatures more frequent and cool summer temperatures less frequent.

Climate change projections

- For the A1B emissions scenario projected increases in temperature range from around 2.5 to 3.5°C over Germany with the agreement between models being moderate over most of the country, but with areas of high agreement in the far south.
- Germany is located in the transition zone between projected increasing precipitation over Northern Europe, and decreases in Southern Europe. Northern parts of Germany are projected to experience increases of up to 10%, whereas the far south has projected decreases of up to 5%. The ensemble agreement over this region is generally moderate, and reflects uncertainty about the precise location of the transition between increasing and decreasing precipitation over Europe.

Climate change impact projections

Crop yields

• Global-scale studies generally project yield increases by the end of the century for wheat and maize, two of Germany's major crops, along with increased land suitability.

However the balance between detrimental ozone effects and CO₂ fertilisation may determine whether crop yield losses or gains are realised under climate change.

• A number of sub-national-scale assessments included here suggest that climate change could be beneficial for the wine and fruits industry in Germany.

Food security

 Germany is currently a country with extremely low levels of undernourishment. Global-scale studies included here generally project that Germany will not face serious food security issues over the next 40 years as a consequence of climate change, especially considering its strong purchasing power and substantial adaptive capacity.

Water stress and drought

- Over 70% of Germany's current water supply is from groundwater. Global-scale studies included here which consider future groundwater withdrawals with climate change suggest that Germany is not particularly vulnerable to increased water stress in the future.
- In contrast, simulations from the AVOID programme project a median increase of around 40% of Germany's population to be exposed to water stress increases by 2100 under SRES A1B. However, the methodology for this study only accounts for changes in run-off, rather than changes in groundwater directly.
- National-scale studies included here suggest that it is possible that droughts could become more frequent with climate change in the east of the country and although annual precipitation may not change, that there could be higher drought risk in summer months.

Pluvial flooding and rainfall

- Large-scale modelling studies support conclusions from the IPCC AR4 that heavy winter precipitation and flooding could increase with climate change for Germany.
- However, recent national-scale assessments suggest that increasing trends in precipitation with climate change for Germany are relatively small when compared with emissions uncertainty and that trends of increasing rainfall with climate change may not be significant.

Fluvial flooding

- Climate change could be associated with increases in fluvial flooding for some parts of Germany, particularly the Rhine basin.
- However, projections of future changes in flood hazard are subject to large uncertainties due to large natural variability as well as large uncertainties in the simulated climate change.
- Simulations from the AVOID programme show a greater tendency towards decreasing flood risk in Germany at first, but later in the century more projections indicate increasing flood risk under the A1B scenario.

Coastal regions

- Germany has coasts on both the North and Baltic Seas extending over five coastal states.
- Large-scale and national-scale studies suggest that without adaptation, the number of people at risk from coastal flooding in Germany due to Sea Level Rise (SLR) under climate change scenarios could be around 300,000. However, these impacts could be reduced substantially by adaptation and climate change mitigation measures.
- One study demonstrates that while Germany is potentially highly threatened by SLR, adaptation in the form of raising of flood dykes and the application of beach nourishment could greatly reduce these impacts to levels which appear manageable.
- These studies add further detail to knowledge documented in the IPCC AR4.

Table of contents

Chapter 1 – Climate Observations	9
Rationale	10
Climate overview Analysis of long-term features in the mean temperature	
Temperature extremes Recent extreme temperature events Heat wave, summer 2003	15
Analysis of long-term features in moderate temperature extremes Attribution of changes in likelihood of occurrence of seasonal mean temperatures Summer 2003	21
Precipitation extremes Recent extreme precipitation events Flooding, August 2002	23
Analysis of long-term features in precipitation	24
Storms Recent storm events Extra-tropical Cyclone Kyrill, January 2007	27
Summary	29
Methodology annex Recent, notable extremes Observational record Analysis of seasonal mean temperature	30 31
Analysis of temperature and precipitation extremes using indices	
Presentation of extremes of temperature and precipitation	
Attribution	46
References	49
Acknowledgements	53
Chapter 2 – Climate Change Projections	55
Introduction	56
Climate projections	58
Summary of temperature change in Germany Summary of precipitation change in Germany	
Chapter 3 – Climate Change Impact Projections	
Introduction Aims and approach Impact sectors considered and methods Supporting literature AVOID programme results Uncertainty in climate change impact assessment	62 62 63 63
Summary of findings for each sector	71
Crop yields	74

Headline Supporting literature Introduction	74
Assessments that include a global or regional perspective	76
National-scale or sub-national scale assessments	81
AVOID programme results Methodology	
Results	
Food security Headline Supporting literature Introduction	
Assessments that include a global or regional perspective	
National-scale or sub-national scale assessments	
Water stress and drought Headline Supporting literature Introduction	
Assessments that include a global or regional perspective	
National-scale or sub-national scale assessments	
AVOID programme results Methodology	
Results	
Pluvial flooding and rainfall	
Headline Supporting literature Introduction	
Assessments that include a global or regional perspective	
National-scale or sub-national scale assessments	
Fluvial flooding Headline Supporting literature Introduction	
Assessments that include a global or regional perspective	
National-scale or sub-national scale assessments	
AVOID programme results Methodology	
Results	
Tropical cyclones	116
Coastal regions Headline Supporting literature Assessments that include a global or regional perspective	
National-scale or sub-national scale assessments	

Chapter 1 – Climate Observations

Rationale

Present day weather and climate play a fundamental role in the day to day running of society. Seasonal phenomena may be advantageous and depended upon for sectors such as farming or tourism. Other events, especially extreme ones, can sometimes have serious negative impacts posing risks to life and infrastructure, and significant cost to the economy. Understanding the frequency and magnitude of these phenomena, when they pose risks or when they can be advantageous and for which sectors of society, can significantly improve societal resilience. In a changing climate it is highly valuable to understand possible future changes in both potentially hazardous events and those reoccurring seasonal events that

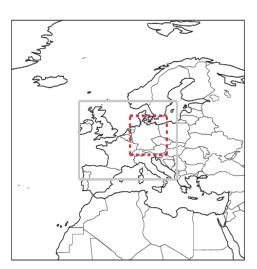


Figure 1. Location of boxes for the regional average time series (red dashed box) in Figures 3 and 5 and the attribution region (grey box) in Figure 4.

are depended upon by sectors such as agriculture and tourism. However, in order to put potential future changes in context, the present day must first be well understood both in terms of common seasonal phenomena and extremes.

The purpose of this chapter is to summarise the weather and climate from 1960 to present day. This begins with a general climate overview including an up to date analysis of changes in surface mean temperature. These changes may be the result of a number of factors including climate change, natural variability and changes in land use. There is then a focus on extremes of temperature, precipitation and storms selected from 2000 onwards, reported in the World Meteorological Organization (WMO) Annual Statement on the Status of the Global Climate and/or the Bulletin of the American Meteorological Society (BAMS) State of the Climate reports. This is followed by a discussion of changes in moderate extremes from 1960 onwards using an updated version of the HadEX extremes database (Alexander et al., 2006) which categorises extremes of temperature and precipitation. These are core climate variables which have received significant effort from the climate research community in terms of data acquisition and processing and for which it is possible to produce long high quality records for monitoring. No new analysis is included for storms (see the methodology annex for background). For seasonal temperature extremes, an attribution analysis then puts the seasons with highlighted extreme events into context of the recent climate versus a

hypothetical climate in the absence of anthropogenic emissions (Christidis et al., 2011). It is important to note that we carry out our attribution analyses on seasonal mean temperatures over the entire country. Therefore these analyses do not attempt to attribute the changed likelihood of individual extreme events. The relationship between extreme events and the large scale mean temperature is likely to be complex, potentially being influenced by *inter alia* circulation changes, a greater expression of natural internal variability at smaller scales, and local processes and feedbacks. Attribution of individual extreme events is an area of developing science. The work presented here is the foundation of future plans to systematically address the region's present and projected future weather and climate, and the associated impacts.

The methodology annex provides details of the data shown here and of the scientific analyses underlying the discussions of changes in the mean temperature and in temperature and precipitation extremes. It also explains the methods used to attribute the likelihood of occurrence of seasonal mean temperatures.

Climate overview

Germany is located between the latitudes of 47° and 55°N, and has a cool temperate climate. Germany has coastlines on both the North Sea and the Baltic Sea, but the coastline is short relative to the size of the country. North Germany is part of the North European Plain and is generally low-lying and rather flat. Central and southern Germany is hilly and rises southwards towards the Alps on the Austrian border. This increasing altitude in the centre and south of the country compensates for the decreasing latitude so that in summer temperatures do not differ much from north to south. However, in winter temperatures decrease, and the number of days with frost and snow increase both southwards and eastwards. This is a consequence of increasing distance from the sea, higher altitude, and the effect being adjacent to the huge land mass of continental Europe.

Rostock, on the Baltic coast, has an annual mean temperature of 9°C, ranging from 1°C in January to 18°C in August. Further south (inland), Berlin is slightly warmer in summer and slightly colder in winter, with an annual mean of 10°C. Further west at Dusseldorf, the annual mean is also 10°C but with less seasonal range. The upper Rhine valley in the south-west is the warmest part of the country in spring and summer, with an annual mean of 11°C at Freiburg, peaking at 20°C in July. Munich, in the south-east at 450m above sea level, has an annual mean temperature of 9°C with cold winters, and the temperature falls further as the altitude increases towards the southern border.

The driest area of Germany is the north-east, with an average annual rainfall as low as 450 mm around Magdeburg and Halle, and 570mm per year at Berlin. This rainfall is well spread through the year, with a slight maximum in summer when the rain is often heavy, thundery and short-lived so that fine, settled weather still dominates. In the west of Germany, rainfall amounts are greater, for example at Dusseldorf which has an annual average amount of 770mm, again with little seasonal variation. Further south rainfall increases, especially during the summer months and with increasing altitude. At Freiburg, the annual average rainfall amount is 930mm and at Munich 970mm.

The weather in Germany is changeable from day to day and the character of individual summers and winters can vary considerably depending on whether oceanic or continental influences dominate. Therefore, hazards such as droughts, floods and heat waves occasionally occur, and cold easterly winds from Russia can bring severe cold spells in the winter.

Analysis of long-term features in the mean temperature

CRUTEM3 data (Brohan et al., 2006) have been used to provide an analysis of mean temperatures from 1960 to 2010 over Germany using the median of pairwise slopes method to fit the trend (Sen, 1968; Lanzante, 1996). The methods are fully described in the methodology annex. In agreement with increasing global average temperatures (Sánchez-Lugo et al., 2011), over the period 1960 to 2010 there is a spatially consistent warming signal for temperature over Germany as shown in Figure 2, consistent with previous research (UNFCCC 2010). Grid boxes in which the 5th to 95th percentiles of the slopes are of the same sign can be more confidently regarded as showing this signal and are geographically widespread for the summer months (June to August). Confidence is lower in the warming signal over winter (December to February). Regionally averaged trends (over grid boxes included in the red dashed box in Figure 1) calculated by the median of pairwise slopes show warming signals with high confidence. This trend for summer is 0.36°C per decade (5th to 95th percentile of slopes: 0.21 to 0.52°C per decade) and for winter is 0.32°C

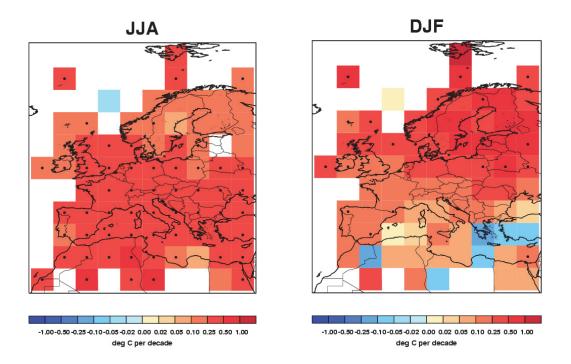


Figure 2. Decadal trends in seasonally averaged temperatures for Germany and the surrounding regions over the period 1960 to 2010. Monthly mean anomalies from CRUTEM3 (Brohan et al., 2006) are averaged over each 3 month season (June-July-August – JJA and December-January-February – DJF). Trends are fitted using the median of pairwise slopes method (Sen, 1968; Lanzante, 1996). There is higher confidence in the trends shown if the 5th to 95th percentiles of the pairwise slopes do not encompass zero because here the trend is considered to be significantly different from a zero trend (no change). This is shown by a black dot in the centre of the respective grid-box.

Temperature extremes

Both hot and cold temperature extremes can place many demands on society. While seasonal changes in temperature are normal and indeed important for a number of societal sectors (e.g. tourism, farming etc.), extreme heat or cold can have serious negative impacts. Importantly, what is 'normal' for one region may be extreme for another region that is less well adapted to such temperatures.

Table 1 shows selected extreme events since 2000 that are reported in WMO Statements on Status of the Global Climate and/or BAMS State of the Climate reports. The heat wave of 2003 is highlighted below as an example of the extreme temperatures which can affect Germany.

Year	Month	Event	Details	Source
2003	Jul-Aug	Heat wave	The 2003 European summer heat wave brought the warmest seasonal temperatures on record to Germany.	WMO (2004)
2006	Jul	Heat wave	Western Europe experienced a summer heat wave; warmest European mean temperature for July	WMO (2007)
2010	Dec	Extreme cold	Coldest December since 1969.	WMO (2011)

Table 1. Selected extreme temperature events reported in WMO Statements on Status of the Global
 Climate and/or BAMS State of the Climate reports since 2000.

Recent extreme temperature events

Heat wave, summer 2003

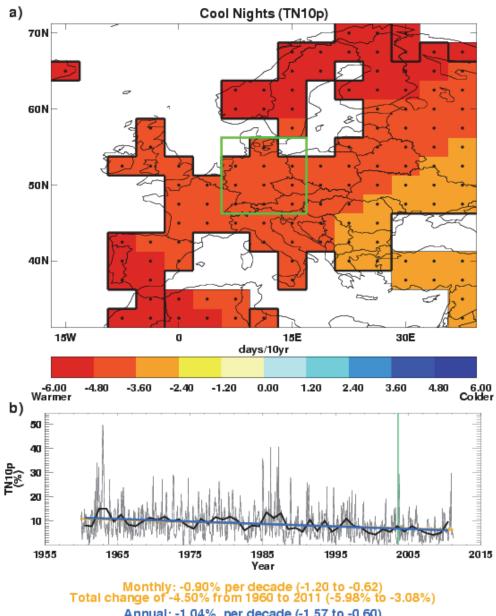
Like most other countries in central and western Europe, Germany experienced a heat wave in early August 2003. The event was associated with persistent high pressure system over Europe that blocked the flow of rain-bearing low-pressure systems (Lyamani et al., 2006). This resulted in the warmest German summer since 1901 at 3.4°C above normal summer temperatures in the 1961 to 1990 period (Levinson and Waple, 2004). Deutsche Wetterdienst daily maximum records show temperatures of 40.2 °C on 9th August in Karlsruhe and again on 13th August in Karlsruhe and in Freiburg (Schönwiese et al., 2004).

Overall, Germany reported the loss of over 9,000 lives due to the extreme weather during August 2003 (EM-DAT, 2011), accounting for 9.6% excess mortality (Sardon, 2007). In the federal state of Baden-Württemberg alone (south-west Germany) 970 to 1,490 deaths can be directly attributed to the 2003 heat wave (WMO, 2004).

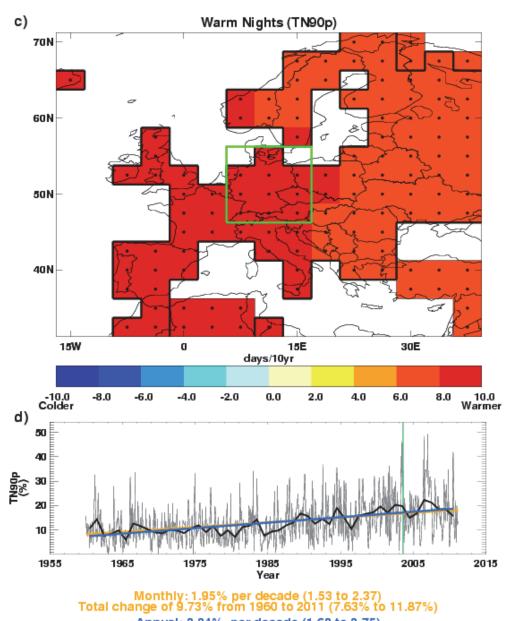
Analysis of long-term features in moderate temperature extremes

ECA&D data (Klein Tank et al., 2002) have been used to update the HadEX extremes analysis for Germany from 1960 to 2010 using daily maximum and minimum temperatures. Here we discuss changes in the frequency of cool days and nights and warm days and nights which are moderate extremes. Cool days/nights are defined as being below the 10th percentile of daily maximum/minimum temperature and warm days/nights are defined as being above the 90th percentile of the daily maximum/minimum temperature. The methods are fully described in the methodology annex.

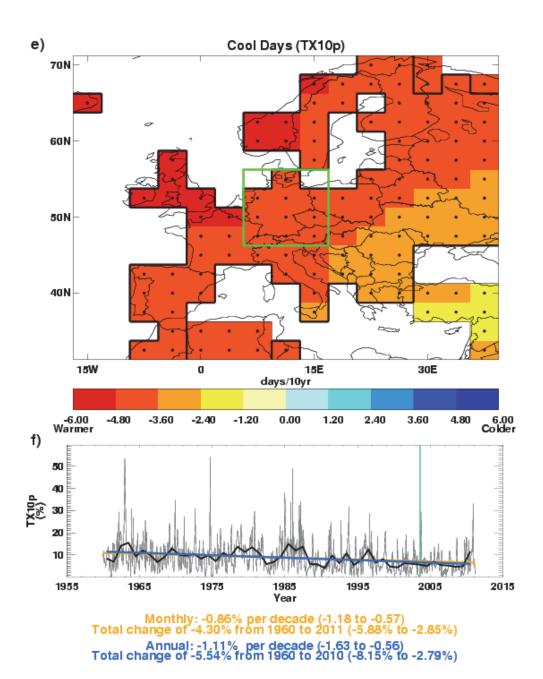
The trends in cool nights and cool days show a substantial decrease across all of Germany with high confidence throughout that the trend is different from zero (Figure 3). There appears to be very little difference between the north and the south of the country. Similarly, there is a strong increase in the numbers of warm nights and warm days over the same period, again with high confidence.











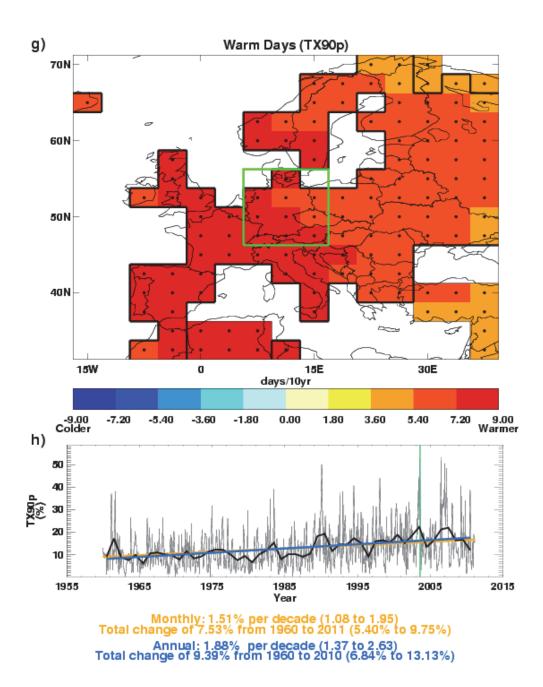


Figure 3. Change in cool nights (a,b), warm nights (c,d), cool days (e,f) and warm days (g,h) for Germany over the period 1960 to 2010 relative to 1961-1990 from the ECA&D dataset (Klein Tank et al., 2002). a,c,e,g) Grid box decadal trends. Grid boxes outlined in solid black contain at least 3 stations and so are likely to be more representative of the wider grid-box. Trends are fitted using the median of pairwise slopes method (Sen, 1968; Lanzante, 1996). Higher confidence in a long-term trend is shown by a black dot if the 5th to 95th percentile slopes are of the same sign. Differences in spatial coverage occur because each index has its own decorrelation length scale (see methodology annex). b,d,f,h) Area averaged annual time series for 5.625° to 16.875° E and 46.25° to 56.25° N as shown by the green box on the map and red box in Figure 1. Thin and thick black lines show the monthly and annual variation respectively. Monthly (orange) and annual (blue) trends are fitted as described above. The decadal trend and its 5th to 95th percentile confidence intervals are stated along with the change over the period for which there are data available. All the trends have higher confidence that they are different from zero as their 5th to 95th percentile slopes are of the same sign. The green vertical lines show the locations of the heat wave in 2003.

The time series plots show a high annual and monthly variability. However, some of the notable temperature extremes of the past few years (highlighted in Table 1) are visible. The heat waves of 2003 and 2006 can clearly be seen.

Attribution of changes in likelihood of occurrence of seasonal mean temperatures

Today's climate covers a range of likely extremes. Recent research has shown that the temperature distribution of seasonal means would likely be different in the absence of anthropogenic emissions (Christidis et al., 2011). Here we discuss the seasonal means, within which the highlighted extreme temperature events occur, in the context of recent climate and the influence of anthropogenic emissions on that climate. The methods are fully described in the methodology annex.

Summer 2003

The distributions of the summer mean regional temperature in recent years in the presence and absence of anthropogenic forcings are shown in Figure 4. Analyses with both models suggest that human influences on the climate have shifted the distribution to higher temperatures. Considering the average over the entire region, the 2003 summer (June-July-August) is exceptionally hot, as it lies at the far end of the warm tail of the temperature distributions for the climate influenced by anthropogenic forcings (red distributions) and is the hottest since 1900 in the CRUTEM3 dataset. In the absence of human influences on the climate (green distributions), the 2003 summer season would have been even more extreme. It should be noted that the attribution results shown here refer to temperature anomalies over the entire region and over an entire season, whereas the actual extreme event had a shorter duration and affected a smaller region.

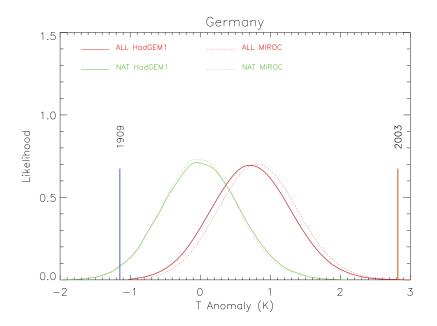


Figure 4. Distributions of the June-July-August mean temperature anomalies (relative to 1961-1990) averaged over a Northern European region that encompasses Germany (10W-20E, 40-60N – as shown in Figure 1) including (red lines) and excluding (green lines) the influence of anthropogenic forcings. The distributions describe the seasonal mean temperatures expected in recent years (2000-2009) and are based on analyses with the HadGEM1 (solid lines) and MIROC (dotted lines) models. The vertical orange and blue lines correspond to the maximum and minimum anomaly in the CRUTEM3 dataset since 1900 respectively.

Precipitation extremes

Precipitation extremes, either excess or deficit, can be hazardous to human health, societal infrastructure, and livestock and agriculture. While seasonal fluctuations in precipitation are normal and indeed important for a number of societal sectors (e.g. tourism, farming etc.), flooding or drought can have serious negative impacts. These are complex phenomena and often the result of accumulated excesses or deficits or other compounding factors such as spring snow-melt, high tides/storm surges or changes in land use. This observation section deals purely with precipitation amounts.

Table 2 shows selected extreme events since 2000 that are reported in WMO Statements on Status of the Global Climate and/or BAMS State of the Climate reports. The floods along the Elbe in August 2002 resulting from an extreme rainfall event is highlighted below as an example of an extreme precipitation event experienced in Germany.

Year	Month	Event	Details	Source
2002	Aug	Flooding	Exceptional rainfall at the beginning of August caused flooding around the Elbe.	WMO (2003)
2006	Apr	Flooding	Melting of large amounts of lying snow coupled with above- average precipitation resulted in one of the largest spring floods in the past 50 years in southern and eastern Germany. The Elbe River rose to record levels in parts of northern Germany, breaking the records set during the 2002 flood.	BAMS (Obregón et al., 2007)

Table 2. Selected extreme precipitation events reported in WMO Statements on Status of the Global
 Climate and/or BAMS State of the Climate reports since 2000.

Recent extreme precipitation events

Flooding, August 2002

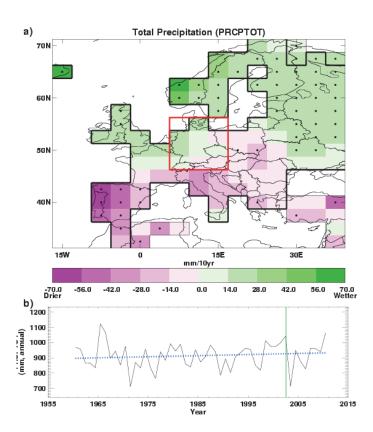
At the beginning of August 2002 an exceptional summer low pressure system caused severe precipitation in the Alps as well as in the Erzgebirge and the Giant Mountains. Zinnwald, located in the eastern part of Germany in the Erzgebirge, endured the heaviest precipitation with 402 mm recorded in 72 hours, and 312 mm in 24 hours on August 12th (DWD, 2011).

312 mm is the highest daily average precipitation ever observed in Germany. The 2002 flood killed 27 people in Germany, affected over 300,000 and is estimated to have cost the German economy \$ 11.6 billion (EM-DAT, 2011).

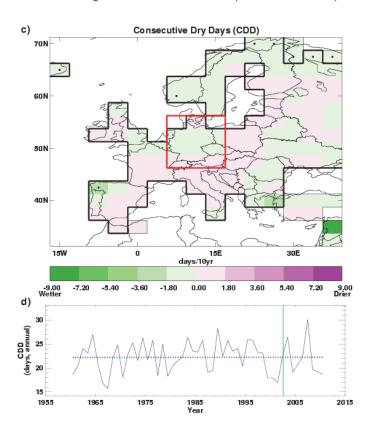
Analysis of long-term features in precipitation

ECA&D data (Klein Tank et al., 2002) have been used to update the HadEX extremes analysis for Germany from 1960 to 2010 for daily precipitation totals. Here we discuss changes in the annual total precipitation, and in the frequency of prolonged (greater than 6 days) wet and dry spells. The methods are fully described in the methodology annex.

The maps of trends in total precipitation, the consecutive wet and consecutive dry days all show a mixed signal over Germany, and there is low confidence that the trends are different from zero (Figure 5). Previous research shows large differences in precipitation trends for Germany. Increases of yearly precipitation are largely limited to the western parts whereas in eastern Germany increases during the winter are largely offset by decreases in the summer (UNFCCC, 2010).



Annual: 7.16mm per decade (-8.40 to 24.87) Total change of 35.82mm from 1960 to 2010 (-42.02mm to 124.37mm)



Annual: 0.01days_per decade (-0.75 to 0.68) Total change of 0.04days from 1960 to 2010 (-3.77days to 3.40days)

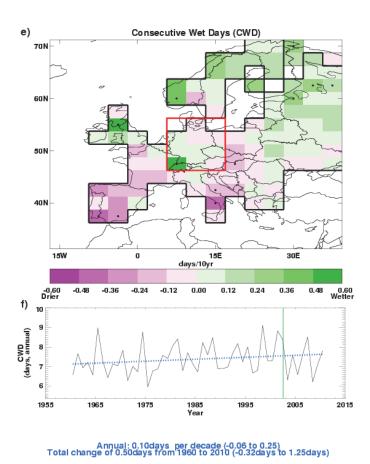


Figure 5. The change in annual total rainfall (a,b), the annual number of continuous dry days (c,d) and the annual number of continuous wet days (e,f) over the period 1960-2010. The maps and time series have been created in exactly the same way as Figure 3. The green vertical lines show the date of the floods of 2002. Only annual regional averages are shown in b,d,f). All the trends have lower confidence that they are different from zero, as their 5th to 95th percentile slopes are of different signs, and hence are marked with dotted lines.

The time series for the precipitation indices also show trends that have low confidence that they are different from zero. The extreme event discussed above does not stand out in the time-series.

Storms

Storms can be very hazardous to all sectors of society. They can be small with localised impacts or spread across wide areas. There is no systematic observational analysis included for storms because, despite recent progress (Peterson et al., 2011; Cornes & Jones, 2011), wind data are not yet adequate for worldwide robust analysis (see methodology annex). Further progress awaits studies of the more reliable barometric pressure data through the new 20th Century Reanalysis (Compo et al., 2011) and its planned successors.

Table 3 shows selected extreme events since 2000 that are reported in WMO Statements on Status of the Global Climate and/or BAMS State of the Climate reports. The winter storm Kyrill in January 2007 is highlighted below as an example of a storm affecting Germany.

Year	Month	Event	Details	Source
2007	Jan	Storm	Winter storm Kyrill.	WMO (2008)
2008	Mar	Storm	Extratropical cyclone Emma. Winds of 160km/h observed in Germany.	WMO (2009) Peterson and Baringer (2009)

Table 3. Selected extreme storm events reported in WMO Statements on Status of the Global

 Climate and/or BAMS State of the Climate reports since 2000.

Recent storm events

Extra-tropical Cyclone Kyrill, January 2007

Low pressure system "Kyrill" was an extratropical cyclone with hurricane-strength winds that formed over Newfoundland on 15th January 2007. It evolved into a very active European windstorm and reached Germany on 18th January (Levinson & Lawrimore, 2008). Wind gusts as high as 202 km/h were recorded on the Wendelstein and 198 km/h in the Harz mountains (DWD, 2007). The media reported the cancellation of many train services (Der Spiegel, 2007a) and the disruption of power to 52,000 homes (Der Spiegel, 2007b). Kyrill

also caused 11 fatalities in Germany (EM-DAT, 2011) the damage caused was estimated at \notin 4.7 billion (Europa, 2007).

Summary

The main features seen in observed climate over Germany from this analysis are:

- Over the period 1960 to 2010 there is a spatially consistent warming signal for temperature over Germany.
- There has been a substantial decrease in the frequency of cool nights and cool days since 1960 and a strong increase in the numbers of warm nights and warm days.
- There has been a general increase in summer temperatures averaged over the country as a result of human influence on climate, making the occurrence of warm summer temperatures more frequent and cool summer temperatures less frequent.

Methodology annex

Recent, notable extremes

In order to identify what is meant by 'recent' events the authors have used the period since 1994, when WMO Status of the Global Climate statements were available to the authors. However, where possible, the most notable events during the last 10 years have been chosen as these are most widely reported in the media, remain closest to the forefront of the memory of the country affected, and provide an example likely to be most relevant to today's society. By 'notable' the authors mean any event which has had significant impact either in terms of cost to the economy, loss of life, or displacement and long term impact on the population. In most cases the events of largest impact on the population have been chosen, however this is not always the case.

Tables of recent, notable extreme events have been provided for each country. These have been compiled using data from the World Meteorological Organisation (WMO) Annual Statements on the Status of the Global Climate. This is a yearly report which includes contributions from all the member countries, and therefore represents a global overview of events that have had importance on a national scale. The report does not claim to capture all events of significance, and consistency across the years of records available is variable. However, this database provides a concise yet broad account of extreme events per country. This data is then supplemented with accounts from the monthly National Oceanic and Atmospheric Administration (NOAA) State of the Climate reports which outline global extreme events of meteorological significance.

We give detailed examples of heat, precipitation and storm extremes for each country where these have had significant impact. Where a country is primarily affected by precipitation or heat extremes this is where our focus has remained. An account of the impact on human life, property and the economy has been given, based largely on media reporting of events, and official reports from aid agencies, governments and meteorological organisations. Some data has also been acquired from the Centre for Research on Epidemiological Disasters (CRED) database on global extreme events. Although media reports are unlikely to be completely accurate, they do give an indication as to the perceived impact of an extreme event, and so are useful in highlighting the events which remain in the national psyche.

Our search for data has not been exhaustive given the number of countries and events included. Although there are a wide variety of sources available, for many events, an official account is not available. Therefore figures given are illustrative of the magnitude of impact only (references are included for further information on sources). It is also apparent that the reporting of extreme events varies widely by region, and we have, where possible, engaged with local scientists to better understand the impact of such events.

The aim of the narrative for each country is to provide a picture of the social and economic vulnerability to the current climate. Examples given may illustrate the impact that any given extreme event may have and the recovery of a country from such an event. This will be important when considering the current trends in climate extremes, and also when examining projected trends in climate over the next century.

Observational record

In this section we outline the data sources which were incorporated into the analysis, the quality control procedure used, and the choices made in the data presentation. As this report is global in scope, including 23 countries, it is important to maintain consistency of methodological approach across the board. For this reason, although detailed datasets of extreme temperatures, precipitation and storm events exist for various countries, it was not possible to obtain and incorporate such a varied mix of data within the timeframe of this project. Attempts were made to obtain regional daily temperature and precipitation data from known contacts within various countries with which to update existing global extremes databases. No analysis of changes in storminess is included as there is no robust historical analysis of global land surface winds or storminess currently available.

Analysis of seasonal mean temperature

Mean temperatures analysed are obtained from the CRUTEM3 global land-based surfacetemperature data-product (Brohan et al., 2006), jointly created by the Met Office Hadley Centre and Climatic Research Unit at the University of East Anglia. CRUTEM3 comprises of more than 4000 weather station records from around the world. These have been averaged together to create 5° by 5° gridded fields with no interpolation over grid boxes that do not contain stations. Seasonal averages were calculated for each grid box for the 1960 to 2010 period and linear trends fitted using the median of pairwise slopes (Sen 1968; Lanzante 1996). This method finds the slopes for all possible pairs of points in the data, and takes their median. This is a robust estimator of the slope which is not sensitive to outlying points. High confidence is assigned to any trend value for which the 5th to 95th percentiles of the pairwise slopes are of the same sign as the trend value and thus inconsistent with a zero trend.

Analysis of temperature and precipitation extremes using indices

In order to study extremes of climate a number of indices have been created to highlight different aspects of severe weather. The set of indices used are those from the World Climate Research Programme (WCRP) Climate Variability and Predictability (CLIVAR) Expert Team on Climate Change Detection and Indices (ETCCDI). These 27 indices use daily rainfall and maximum and minimum temperature data to find the annual (and for a subset of the indices, monthly) values for, e.g., the 'warm' days where daily maximum temperature exceeds the 90th percentile maximum temperature as defined over a 1961 to 1990 base period. For a full list of the indices we refer to the website of the ETCCDI (http://cccma.seos.uvic.ca/ETCCDI/index.shtml).

Index	Description	Shortname	Notes
Cool night frequency	Daily minimum temperatures lower than the 10 th percentile daily minimum temperature using the base reference period 1961-1990	TN10p	
Warm night frequency	Daily minimum temperatures higher than the 90 th percentile daily minimum temperature using the base reference period 1961-1990	TN90p	
Cool day frequency	Daily maximum temperatures lower than the 10 th percentile daily maximum temperature using the base reference period 1961-1990	TX10p	
Warm day frequency	Daily maximum temperatures higher than the 90 th percentile daily maximum temperature using the base reference period 1961-1990	ТХ90р	
Dry spell duration	Maximum duration of continuous days within a year with rainfall <1mm	CDD	Lower data coverage due to the requirement for a 'dry spell' to be at least 6 days long resulting in intermittent temporal coverage
Wet spell duration	Maximum duration of continuous days with rainfall >1mm for a given year	CWD	Lower data coverage due to the requirement for a 'wet spell' to be at least 6 days long resulting in intermittent temporal coverage
Total annual precipitation	Total rainfall per year	PRCPTOT	

Table 4. Description of ETCCDI indices used in this document.

A previous global study of the change in these indices, containing data from 1951-2003 can be found in Alexander et al. 2006, (HadEX; see http://www.metoffice.gov.uk/hadobs/hadex/). In this work we aimed to update this analysis to the present day where possible, using the most recently available data. A subset of the indices is used here because they are most easily related to extreme climate events (Table 4).

Use of HadEX for analysis of extremes

The HadEX dataset comprises all 27 ETCCDI indices calculated from station data and then smoothed and gridded onto a $2.5^{\circ} \times 3.75^{\circ}$ grid, chosen to match the output from the Hadley Centre suite of climate models. To update the dataset to the present day, indices are

calculated from the individual station data using the RClimDex/FClimDex software; developed and maintained on behalf of the ETCCDI by the Climate Research Branch of the Meteorological Service of Canada. Given the timeframe of this project it was not possible to obtain sufficient station data to create updated HadEX indices to present day for a number of countries: Brazil; Egypt; Indonesia; Japan (precipitation only); South Africa; Saudi Arabia; Peru; Turkey; and Kenya. Indices from the original HadEX data-product are used here to show changes in extremes of temperature and precipitation from 1960 to 2003. In some cases the data end prior to 2003. Table 5 summarises the data used for each country. Below, we give a short summary of the methods used to create the HadEX dataset (for a full description see Alexander et al., 2006).

To account for the uneven spatial coverage when creating the HadEX dataset, the indices for each station were gridded, and a land-sea mask from the HadCM3 model applied. The interpolation method used in the gridding process uses a decorrelation length scale (DLS) to determine which stations can influence the value of a given grid box. This DLS is calculated from the e-folding distance of the individual station correlations. The DLS is calculated separately for five latitude bands, and then linearly interpolated between the bands. There is a noticeable difference in spatial coverage between the indices due to these differences in decorrelation length scales. This means that there will be some grid-box data where in fact there are no stations underlying it. Here we apply black borders to grid-boxes where at least 3 stations are present to denote greater confidence in representation of the wider grid-box area there. The land-sea mask enables the dataset to be used directly for model comparison with output from HadCM3. It does mean, however, that some coastal regions and islands over which one may expect to find a grid-box are in fact empty because they have been treated as sea

Data Sources used for updates to the HadEX analysis of extremes

We use a number of different data sources to provide sufficient coverage to update as many countries as possible to present day. These are summarised in Table 5. In building the new datasets we have tried to use exactly the same methodology as was used to create the original HadEX to retain consistency with a product that was created through substantial international effort and widely used, but there are some differences, which are described in the next section.

Wherever new data have been used, the geographical distributions of the trends were compared to those obtained from HadEX, using the same grid size, time span and fitting method. If the pattern of the trends in the temperature or precipitation indices did not match that from HadEX, we used the HadEX data despite its generally shorter time span. Differences in the patterns of the trends in the indices can arise because the individual stations used to create the gridded results are different from those in HadEX, and the quality control procedures used are also very likely to be different. Countries where we decided to use HadEX data despite the existence of more recent data are Egypt and Turkey.

GHCND:

The Global Historical Climate Network Daily data has near-global coverage. However, to ensure consistency with the HadEX database, the GHCND stations were compared to those stations in HadEX. We selected those stations which are within 1500m of the stations used in the HadEX database and have a high correlation with the HadEX stations. We only took the precipitation data if its r>0.9 and the temperature data if one of its r-values >0.9. In addition, we required at least 5 years of data beyond 2000. These daily data were then converted to the indices using the *fclimdex* software.

ECA&D and SACA&D:

The European Climate Assessment and Dataset and the Southeast Asian Climate Assessment and Dataset data are pre-calculated indices comprising the core 27 indices from the ETCCDI as well as some extra ones. We kindly acknowledge the help of Albert Klein Tank, the KNMI¹ and the BMKG² for their assistance in obtaining these data.

Mexico:

The station data from Mexico has been kindly supplied by the SMN³ and Jorge Vazquez. These daily data were then converted to the required indices using the *Fclimdex* software. There are a total of 5298 Mexican stations in the database. In order to select those which have sufficiently long data records and are likely to be the most reliable ones we performed a cross correlation between all stations. We selected those which had at least 20 years of

¹ Koninklijk Nederlands Meteorologisch Instituut – The Royal Netherlands Meteorological Institute

² Badan Meteorologi, Klimatologi dan Geofisika – The Indonesian Meteorological, Climatological and Geophysical Agency

³ Servicio Meteorológico Nacional de México – The Mexican National Meteorological Service

data post 1960 and have a correlation with at least one other station with an r-value >0.95. This resulted in 237 stations being selected for further processing and analysis.

Indian Gridded:

The India Meteorological Department provided daily gridded data (precipitation 1951-2007, temperature 1969-2009) on a $1^{\circ} \times 1^{\circ}$ grid. These are the only gridded daily data in our analysis. In order to process these in as similar a way as possible the values for each grid were assumed to be analogous to a station located at the centre of the grid. We keep these data separate from the rest of the study, which is particularly important when calculating the decorrelation length scale, which is on the whole larger for these gridded data.

	1				1		
Notes		Land-sea mask has been adapted to include Tasmania and the area around Brisbane	Interpolated from Indian Gridded data	Spatial coverage is poor	Precipitation has very poor coverage beyond 1997 except in 2003-04, and no data at all in 2000-02, 2005-11	There are no data for Egypt so all grid- box values have been interpolated from stations in Jordan, Israel, Libya and Sudan	
Temporal resolution available	annual	monthly, seasonal and annual	monthly, seasonal and annual	annual	monthly, seasonal and annual	annual	monthly, seasonal and annual
Indices included (see Table 1 for details)	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT,	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD
Period of data coverage (T = temperature, P = precipitation)	1960-2010 (T,P)	1960-2010 (T,P)	1960-2007 (P), 1970-2009 (T)	1960-2000 (P) 2002 (T)	1960-1997 (P) 1960-2003 (T _{min}) 1960-2010 (T _{max})	No data	1960-2010 (T,P)
Data source (T = temperature, P = precipitation)	Matilde Rusticucci (T,P)	GHCND (T,P)	Indian Gridded data (T,P)	HadEX (T,P)	GHCND (T,P)	HadEX (T,P)	ECA&D (T,P)
Region box (red dashed boxes in Fig. 1 and on each map at beginning of chapter)	73.125 to 54.375 ° W, 21.25 to 56.25 ° S	114.375 to 155.625 ° E, 11.25 to 43.75 ° S	88.125 to 91.875 ° E, 21.25 to 26.25 ° N	73.125 to 31.875 ° W, 6.25 ° N to 33.75 ° S	73.125 to 133.125 ° E, 21.25 to 53.75 ° N	24.375 to 35.625 ° E, 21.25 to 31.25 ° N	5.625 ° W to 9.375 ° E, 41.25 to 51.25 ° N
Country	Argentina	Australia	Bangladesh	Brazil	China	Egypt	France

37

		Spatial coverage is poor	Land-sea mask has been adapted to improve coverage of Italy		There are no temperature data for Kenya and so grid-box values have been interpolated from neighbouring Uganda and the United Republic of Tanzania. Regional averages include grid-boxes from outside Kenya that enable continuation to 2003	237/5298 stations selected. Non uniform spatial coverage. Drop in T and P coverage in 2009.	Intermittent coverage in TX90p, CDD and CWD
monthly, seasonal and annual	monthly, seasonal and annual	annual	monthly, seasonal and annual	monthly, seasonal and annual (T), annual (P)	annual	monthly, seasonal and annual	annual
TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT,	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT,	TN10p, TN90p, TX10p, TX90p, PRCPTOT	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD
1960-2010 (T,P)	1960-2003 (P), 1970-2009 (T)	1968-2003 (T,P)	1960-2010 (T,P)	1960-2003 (P) 1960-2000 (T _{min}) 1960-2010 (T _{max})	1960-1999 (P)	1960-2009 (T,P)	1960-2002 (T,P)
ECA&D (T,P)	Indian Gridded data (T,P)	HadEX (T,P)	ECA&D (T,P)	HadEX (P) GHCND (T)	HadEX (T,P)	Raw station data from the Servicio Meteorológico Nacional (SMN) (T,P)	HadEX (T,P)
5.625 to 16.875 ° E, 46.25 to 56.25 ° N	69.375 to 99.375 ° E, 6.25 to 36.25 ° N	95.625 to 140.625 ° E, 6.25 ° N to 11.25 ° S	5.625 to 16.875 ° E, 36.25 to 46.25 ° N	129.375 to 144.375 ° E, 31.25 to 46.25 ° N	31.875 to 43.125 ° E, 6.25 ° N to 6.25 ° S	118.125 to 88.125 ° W, 13.75 to 33.75 ° N	84.735 to 65.625 ° W, 1.25 ° N to 18.75 ° S
Germany	India	Indonesia	Italy	Japan	Kenya	Mexico	Peru

38

Country split for presentation purposes only.	Spatial coverage is poor		There are too few data points for CWD to calculate trends or regional timeseries		Intermittent coverage in CWD and CDD with no regional average beyond 2000		
monthly, seasonal and annual	annual	annual	annual	monthly, seasonal and annual	annual	monthly, seasonal and annual	monthly, seasonal and annual
TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD	TN10p, TN90p, TX10p, TX90p, PRCPTOT, CDD, CWD
1960-2010 (T,P)	1960-2000 (T,P)	1960-2000 (T,P)	1960-2003 (T,P)	1960-2010 (T,P)	1960-2003 (T,P)	1960-2010 (T,P)	1960-2010 (T,P)
ECA&D (T,P)	HadEX (T,P)	HadEX (T,P)	HadEX (T,P)	ECA&D (T,P)	HadEX (T,P)	ECA&D (T,P)	GHCND (T,P)
West Russia 28.125 to 106.875 ° E, 43.75 to 78.75 ° N, East Russia 103.125 to 189.375 ° E, 43.75 to 78.75 ° N	31.875 to 54.375 ° E, 16.25 to 33.75 ° N	13.125 to 35.625 ° W, 21.25 to 36.25 ° S	125.625 to 129.375 ° E, 33.75 to 38.75 ° N	9.375 ° W to 1.875 ° E, 36.25 to 43.75 ° N	24.375 to 46.875 ° E, 36.25 to 43.75 ° N	9.375 ° W to 1.875 ° E, 51.25 to 58.75 ° N	125.625 to 65.625 ° W, 23.75 to 48.75 ° N
Russia	Saudi Arabia	South Africa	Republic of Korea	Spain	Turkey	United Kingdom	United States of America

Table 5. Summary of data used for each country.

Quality control and gridding procedure used for updates to the HadEX analysis of extremes

In order to perform some basic quality control checks on the index data, we used a two-step process on the indices. Firstly, internal checks were carried out, to remove cases where the 5 day rainfall value is less than the 1 day rainfall value, the minimum T_min is greater than the minimum T_max and the maximum T_min is greater than the maximum T_max. Although these are physically impossible, they could arise from transcription errors when creating the daily dataset, for example, a misplaced minus sign, an extra digit appearing in the record or a column transposition during digitisation. During these tests we also require that there are at least 20 years of data in the period of record for the index for that station, and that some data is found in each decade between 1961 and 1990, to allow a reasonable estimation of the climatology over that period.

Weather conditions are often similar over many tens of kilometres and the indices calculated in this work are even more coherent. The correlation coefficient between each station-pair combination in all the data obtained is calculated for each index (and month where appropriate), and plotted as a function of the separation. An exponential decay curve is fitted to the data, and the distance at which this curve has fallen by a factor 1/*e* is taken as the decorrelation length scale (DLS). A DLS is calculated for each dataset separately. For the GHCND, a separate DLS is calculated for each hemisphere. We do not force the fitted decay curve to show perfect correlation at zero distance, which is different to the method employed when creating HadEX. For some of the indices in some countries, no clear decay pattern was observed in some data sets or the decay was so slow that no value for the DLS could be determined. In these cases a default value of 200km was used.

We then perform external checks on the index data by comparing the value for each station with that of its neighbours. As the station values are correlated, it is therefore likely that if one station measures a high value for an index for a given month, its neighbours will also be measuring high. We exploit this coherence to find further bad values or stations as follows. Although raw precipitation data shows a high degree of localisation, using indices which have monthly or annual resolution improves the coherence across wider areas and so this neighbour checking technique is a valid method of finding anomalous stations.

We calculate a climatology for each station (and month if appropriate) using the mean value for each index over the period 1961-1990. The values for each station are then anomalised using this climatology by subtracting this mean value from the true values, so that it is clear if the station values are higher or lower than normal. This means that we do not need to take

differences in elevation or topography into account when comparing neighbours, as we are not comparing actual values, but rather deviations from the mean value.

All stations which are within the DLS distance are investigated and their anomalised values noted. We then calculate the weighted median value from these stations to take into account the decay in the correlation with increasing distance. We use the median to reduce the sensitivity to outliers.

If the station value is greater than 7.5 median-absolute-deviations away from the weighted median value (this corresponds to about 5 standard deviations if the distribution is Gaussian, but is a robust measure of the spread of the distribution), then there is low confidence in the veracity of this value and so it is removed from the data.

To present the data, the individual stations are gridded on a 3.75° x 2.5° grid, matching the output from HadCM3. To determine the value of each grid box, the DLS is used to calculate which stations can reasonably contribute to the value. The value of each station is then weighted using the DLS to obtain a final grid box value. At least three stations need to have valid data and be near enough (within 1 DLS of the gridbox centre) to contribute in order for a value to be calculated for the grid point. As for the original HadEX, the HadCM3 land-sea mask is used. However, in three cases the mask has been adjusted as there are data over Tasmania, eastern Australia and Italy that would not be included otherwise (Figure 6).

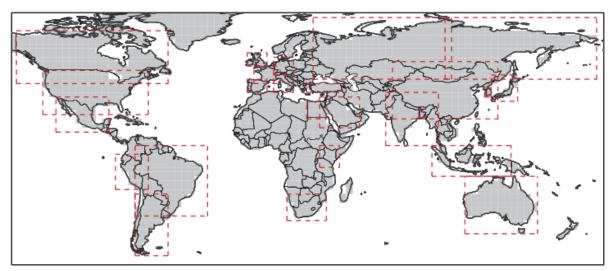


Figure 6. Land-sea mask used for gridding the station data and regional areas allocated to each country as described in Table 5.

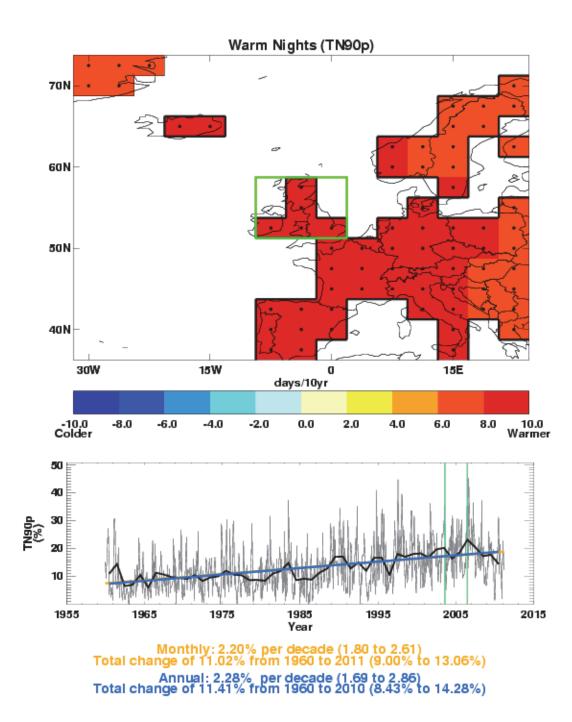
Presentation of extremes of temperature and precipitation

Indices are displayed as regional gridded maps of decadal trends and regional average timeseries with decadal trends where appropriate. Trends are fitted using the median of pairwise slopes method (Sen 1968, Lanzante 1996). Trends are considered to be significantly different from a zero trend if the 5th to 95th percentiles of the pairwise slopes do not encompass zero. This is shown by a black dot in the centre of the grid-box or by a solid line on time-series plots. This infers that there is high confidence in the sign (positive or negative) of the sign. Confidence in the trend magnitude can be inferred by the spread of the 5th to 95th percentiles of the pairwise slopes which is given for the regional average decadal trends. Trends are only calculated when there are data present for at least 50% of years in the period of record and for the updated data (not HadEX) there must be at least one year in each decade.

Due to the practice of data-interpolation during the gridding stage (using the DLS) there are values for some grid boxes when no actually station lies within the grid box. There is more confidence in grid boxes for which there are underlying data. For this reason, we identify those grid boxes which contain at least 3 stations by a black contour line on the maps. The DLS differs with region, season and index which leads to large differences in the spatial coverage. The indices, by their nature of being largely threshold driven, can be intermittent over time which also effects spatial and temporal coverage (see Table 4).

Each index (and each month for the indices for which there is monthly data) has a different DLS, and so the coverage between different indices and datasets can be different. The restrictions on having at least 20 years of data present for each input station, at least 50% of years in the period of record and at least one year in each decade for the trending calculation, combined with the DLS, can restrict the coverage to only those regions with a dense station network reporting reliably.

Each country has a rectangular region assigned as shown by the red dashed box on the map in Figure 1 and listed in Table 2, which is used for the creation of the regional average. This is sometimes identical to the attribution region shown in grey on the map in Figure 1. This region is again shown on the maps accompanying the time series of the regional averages as a reminder of the region and grid boxes used in the calculation. Regional averages are created by weighting grid box values by the cosine of their grid box centre latitude. To ensure consistency over time a regional average is only calculated when there are a sufficient number of grid boxes present. The full-period median number of grid-boxes present is calculated. For regions with a median of more than six grid-boxes there must be at least 80% of the median number of grid boxes present for any one year to calculate a regional average. For regions with six or fewer median grid boxes this is relaxed to 50%. These limitations ensure that a single station or grid box which has a longer period of record than its neighbours cannot skew the timeseries trend. So sometimes there may be grid-boxes present but no regional average time series. The trends for the regional averages are calculated in the same way as for the individual grid boxes, using the median of pairwise slopes method (Sen 1968, Lanzante 1996). Confidence in the trend is also determined if the 5th to 95th percentiles of the pairwise slopes are of the same sign and thus inconsistent with a zero trend. As well as the trend in quantity per decade, we also show the full change in the quantity from 1960 to 2010 that this fitted linear trend implies.



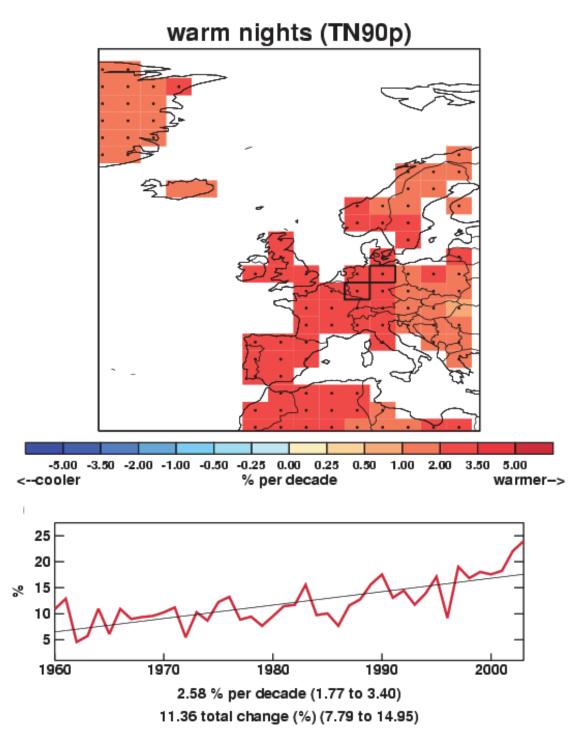


Figure 7. Examples of the plots shown in the data section. Left: From ECA&D data between 1960-2010 for the number of warm nights, and Right: from HadEX data (1960-2003) for the total precipitation. A full explanation of the plots is given in the text below.

The results are presented in the form of a map and a time series for each country and index. The map shows the grid box decadal trend in the index over the period for which there are data. High confidence, as determined above, is shown by a black dot in the grid box centre. To show the variation over time, the values for each year (and month if available) are shown in a time series for a regional average. The values of the indices have been normalised to a base period of 1961-1990 (except the Indian gridded data which use a 1971 to 1990 period), both in HadEX and in the new data acquired for this project. Therefore, for example, the percentage of nights exceeding the 90th percentile for a temperature is 10% for that period.

There are two influences on whether a grid box contains a value or not – the land-sea mask, and the decorrelation length scale. The land-sea mask is shown in Figure 6. There are grid boxes which contain some land but are mostly sea and so are not considered. The decorrelation length scale sets the maximum distance a grid box can be from stations before no value is assigned to it. Grid boxes containing three or more stations are highlighted by a thick border. This indicates regions where the value shown is likely to be more representative of the grid box area mean as opposed to a single station location.

On the maps for the new data there is a box indicating which grid boxes have been extracted to calculate the area average for the time series. This box is the same as shown in Figure 1 at the beginning of each country's document. These selected grid boxes are combined using area (cosine) weighting to calculate the regional average (both annual [thick lines] and monthly [thin lines] where available). Monthly (orange) and annual (blue) trends are fitted to these time series using the method described above. The decadal trend and total change over the period where there are data are shown with 5th to 95th percentile confidence intervals in parentheses. High confidence, as determined above, is shown by a solid line as opposed to a dotted one. The green vertical lines on the time series show the dates of some of the notable events outlined in each section.

Attribution

Regional distributions of seasonal mean temperatures in the 2000s are computed with and without the effect of anthropogenic influences on the climate. The analysis considers temperatures averaged over the regions shown in Figure 8. These are also identified as grey boxes on the maps in Figure 1. The coordinates of the regions are given in Table 6. The methodology combines information from observations and model simulations using the approach originally introduced in Christidis et al., 2010 and later extended in Christidis et al., 2011, where more details can be found. The analysis requires spatial scales greater than about 2,500 km and for that reason the selected regions (Fig.8 and Table 6) are often larger than individual countries, or include several smaller countries in a single region (for example UK, Germany and France are grouped in one region).

Observations of land temperature come from the CRUTEM3 gridded dataset (Brohan et al., 2006) and model simulations from two coupled GCMs, namely the Hadley Centre HadGEM1 model (Martin et al., 2006) and version 3.2 of the MIROC model (K-1 Developers, 2004). The use of two GCMs helps investigate the sensitivity of the results to the model used in the analysis. Ensembles of model simulations from two types of experiments are used to partition the temperature response to external forcings between its anthropogenic and natural components. The first experiment (ALL) simulates the combined effect of natural and anthropogenic forcings on the climate system and the second (ANTHRO) includes anthropogenic forcings only. The difference of the two gives an estimate of the effect of the natural forcings (NAT). Estimates of the effect of internal climate variability are derived from long control simulations of the unforced climate. Distributions of the regional summer mean temperature are computed as follows:

- a) A global optimal fingerprinting analysis (Allen and Tett, 1999; Allen and Stott, 2003) is first carried out that scales the global simulated patterns (fingerprints) of climate change attributed to different combinations of external forcings to best match them to the observations. The uncertainty in the scaling that originates from internal variability leads to samples of the scaled fingerprints, i.e. several realisations that are plausibly consistent with the observations. The 2000-2009 decade is then extracted from the scaled patterns and two samples of the decadal mean temperature averaged over the reference region are then computed with and without human influences, which provide the Probability Density Functions (PDFs) of the decadal mean temperature attributable to ALL and NAT forcings.
- b) Model-derived estimates of noise are added to the distributions to take into account the uncertainty in the simulated fingerprints.
- c) In the same way, additional noise from control model simulations is introduced to the distributions to represent the effect of internal variability in the annual values of the seasonal mean temperatures. The result is a pair of estimated distributions of the annual values of the seasonal mean temperature in the region with and without the effect of human activity on the climate. The temperatures throughout the analysis are expressed as anomalies relative to period 1961-1990.

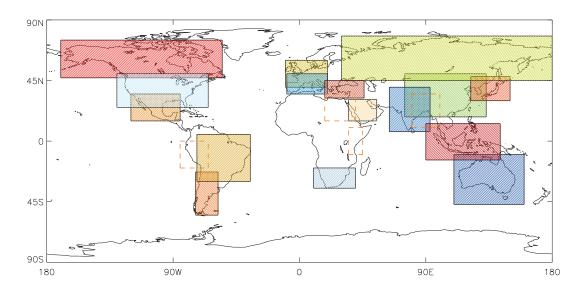


Figure 8. The regions used in the attribution analysis. Regions marked with dashed orange boundaries correspond to non-G20 countries that were also included in the analysis.

Region	Region Coordinates
Argentina	74-58W, 55-23S
Australia	110-160E, 47-10S
Bangladesh	80-100E, 10-35N
Brazil	73-35W, 30S-5N
Canada-Alaska	170-55W, 47-75N
China	75-133E, 18-50N
Egypt	18-40E, 15-35N
France-Germany-UK	10W-20E, 40-60N
India	64-93E, 7-40N
Indonesia	90-143E, 14S-13N
Italy-Spain	9W-20E, 35-50N
Japan-Republic of Korea	122-150E, 30-48N
Kenya	35-45E, 10S-10N
Mexico	120-85W, 15-35N
Peru	85-65W, 20-0S
Russia	30-185E, 45-78N
Saudi Arabia	35-55E, 15-31N
South Africa	10-40E, 35-20S
Turkey	18-46E, 32-45N

Table 6. The coordinates of the regions used in the attribution analysis.

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Acknowledgements

Data for this work were taken from the ECA&D dataset (Klein Tank et al., 2002). We thank Lisa Alexander and Markus Donat (University of New South Wales) for their help and advice. We also thank reviewers from Germany for their valuable input and advice.

Chapter 2 – Climate Change Projections

Introduction

Climate models are used to understand how the climate will evolve over time and typically represent the atmosphere, ocean, land surface, cryosphere, and biogeochemical processes, and solve the equations governing their evolution on a geographical grid covering the globe. Some processes are represented explicitly within climate models, large-scale circulations for instance, while others are represented by simplified parameterisations. The use of these parameterisations is sometimes due to processes taking place on scales smaller than the typical grid size of a climate model (a Global Climate Model (GCM) has a typical horizontal resolution of between 250 and 600km) or sometimes to the current limited understanding of these processes. Different climate modelling institutions use different plausible representations of the climate system, which is why climate projections for a single greenhouse gas emissions scenario differ between modelling institutes. This gives rise to "climate model structural uncertainty".

In response to a proposed activity of the World Climate Research Programme's (WCRP's; http://www.wcrp-climate.org/) Working Group on Coupled Modelling (WGCM), the Program for Climate Model Diagnosis and Intercomparison (PCMDI; http://www-pcmdi.llnl.gov/) volunteered to collect model output contributed by leading climate modelling centres around the world. Climate model output from simulations of the past, present and future climate was collected by PCMDI mostly during the years 2005 and 2006, and this archived data constitutes phase 3 of the Coupled Model Intercomparison Project (CMIP3). In part, the WGCM organised this activity to enable those outside the major modelling centres to perform research of relevance to climate scientists preparing the IPCC Fourth Assessment Report (AR4). This unprecedented collection of recent model output is commonly known as the "CMIP3 multi-model dataset". The GCMs included in this dataset are referred to regularly throughout this review, although not exclusively.

The CMIP3 multi-model ensemble has been widely used in studies of regional climate change and associated impacts. Each of the constituent models was subject to extensive testing by the contributing institute, and the ensemble has the advantage of having been constructed from a large pool of alternative model components, therefore sampling alternative structural assumptions in how best to represent the physical climate system. Being assembled on an opportunity basis, however, the CMIP3 ensemble was not designed to represent model uncertainties in a systematic manner, so it does not, in isolation, support

robust estimates of the risk of different levels of future climate change, especially at a regional level.

Since CMIP3, a new (CMIP5) generation of coupled ocean-atmosphere models has been developed, which is only just beginning to be available and is being used for new projections for the IPCC Fifth Assessment Report (AR5).

These newer models typically feature higher spatial resolution than their CMIP3 counterparts, including in some models a more realistic representation of stratosphere-troposphere interactions. The CMIP5 models also benefit from several years of development in their parameterisations of small scale processes, which, together with resolution increases, are expected to result in a general improvement in the accuracy of their simulations of historical climate, and in the credibility of their projections of future changes. The CMIP5 programme also includes a number of comprehensive Earth System Models (ESMs) which explicitly simulate the earth's carbon cycle and key aspects of atmospheric chemistry, and also contain more sophisticated representations of aerosols compared to CMIP3 models.

The CMIP3 results should be interpreted as a useful interim set of plausible outcomes. However, their neglect of uncertainties, for instance in carbon cycle feedbacks, implies that higher levels of warming outside the CMIP3 envelope cannot be ruled out. In future, CMIP5 coupled model and ESM projections can be expected to produce improved advice on future regional changes. In particular, ensembles of ESM projections will be needed to provide a more comprehensive survey of possible future changes and their relative likelihoods of occurrence. This is likely to require analysis of the CMIP5 multi-model ESM projections, augmented by larger ensembles of ESM simulations in which uncertainties in physical and biogeochemical feedback processes can be explored more systematically, for example via ensembles of model runs in which key aspects of the climate model are slightly adjusted. Note that such an exercise might lead to the specification of wider rather than narrower uncertainties compared to CMIP3 results, if the effects of representing a wider range of earth system processes outweigh the effects of refinements in the simulation of physical atmosphere-ocean processes already included in the CMIP3 models.

Climate projections

The Met Office Hadley Centre is currently producing perturbed parameter ensembles of a single model configuration known as HadCM3C, to explore uncertainties in physical and biogeochemical feedback processes. The results of this analysis will become available in the next year and will supplement the CMIP5 multi-model ESM projections, providing a more comprehensive set of data to help progress understanding of future climate change. However, many of the studies covered in the chapter on climate impacts have used CMIP3 model output. For this reason, and because it is still the most widely used set of projections available, the CMIP3 ensemble output for temperature and precipitation, for the A1B emission scenario, for Germany and the surrounding region is shown below.

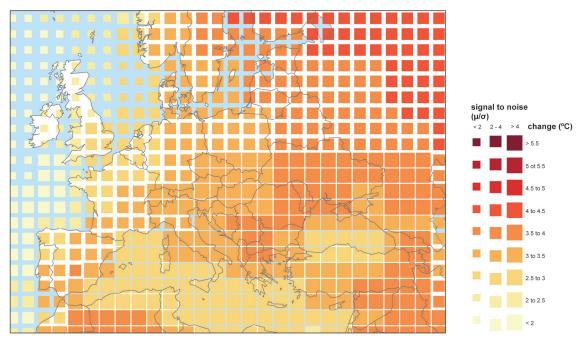


Figure 1. Percentage change in average annual temperature by 2100 from 1960-1990 baseline climate, averaged over 21 CMIP3 models. The size of each pixel represents the level of agreement between models on the magnitude of the change.

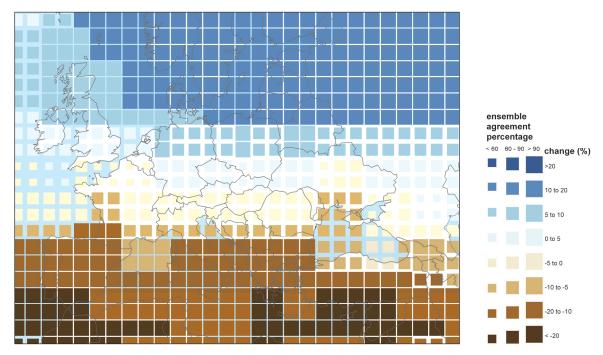


Figure 2. Percentage change in average annual precipitation by 2100 from 1960-1990 baseline climate, averaged over 21 CMIP3 models. The size of each pixel represents the level of agreement between models on the sign of the change.

Summary of temperature change in Germany

Figure 1 shows the percentage change in average annual temperature by 2100 from 1960-1990 baseline climate, averaged over 21 CMIP3 models. All of the models in the CMIP3 ensemble project increased temperatures in the future, but the size of each pixel indicates how well the models agree over the magnitude of the increase.

Projected increases in temperature range from around 2.5 to 3.5°C over Germany with the agreement between models being moderate over most of the country, but with areas of high agreement in the far south.

Summary of precipitation change in Germany

Figure 2 shows the percentage change in average annual precipitation by 2100 from 1960-1990 baseline climate, averaged over 21 CMIP3 models. Unlike for temperature, the models sometimes disagree over whether precipitation is increasing or decreasing over a region, so in this case the size of each pixel indicates the percentage of the models in the ensemble that agree on the sign of the change in precipitation.

Germany is located in the transition zone between projected increasing precipitation over Northern Europe, and decreases in Southern Europe. Northern parts of Germany are projected to experience increases of up to 10%, whereas the far south has projected decreases of up to 5%. The ensemble agreement over this region is generally moderate, and reflects uncertainty about the precise location of the transition between increasing and decreasing precipitation over Europe.

Chapter 3 – Climate Change Impact Projections

Introduction

Aims and approach

This chapter looks at research on a range of projected climate change impacts, with focus on results for Germany. It includes projections taken from the AVOID programme, for some of the impact sectors.

The aim of this work is to take a 'top down' approach to assessing global impacts studies, both from the literature and from new research undertaken by the AVOID programme. This project covers 23 countries, with summaries from global studies provided for each of these. This global approach allows some level of comparison between countries, whilst presenting information on a scale most meaningful to inform international policy.

The literature covered in this chapter focuses on research published since the Fourth Assessment Report (AR4) of the Intergovernmental Panel on Climate Change (IPCC) and should be read in conjunction with IPCC AR4 WG1 and WG2 reports. For some sectors considered, an absence of research developments since the IPCC AR4, means earlier work is cited as this helps describe the current level of scientific understanding. This report focuses on assessing scientific research about climate change impacts within sectors; it does not present an integrated analysis of climate change adaptation policies.

Some national and sub-national scale literature is reported to a limited extent to provide some regional context.

Impact sectors considered and methods

This report reviews the evidence for the impact of climate change on a number of sectors, for Germany. The following sectors are considered in turn in this report:

- Crop yields
- Food security
- Water stress and drought
- Pluvial flooding and rainfall

- Fluvial flooding
- Tropical cyclones (where applicable)
- Coastal regions

Supporting literature

Literature searches were conducted for each sector with the Thomson Reuters Web of Science (WoS., 2011) and Google Scholar academic search engines respectively. Furthermore, climate change impact experts from each of the 23 countries reviewed were contacted. These experts were selected through a combination of government nomination and from experts known to the Met Office. They were asked to provide literature that they felt would be of relevance to this review. Where appropriate, such evidence has been included. A wide range of evidence was considered, including; research from international peer-reviewed journal papers; reports from governments, non-governmental organisations, and private businesses (e.g. reinsurance companies), and research papers published in national journals.

For each impact sector, results from assessments that include a global- or regional-scale perspective are considered separately from research that has been conducted at the national- or sub-national-scale. The consideration of global- and regional-scale studies facilitates a comparison of impacts across different countries, because such studies apply a consistent methodology for each country. While results from national- and sub-national-scale studies are not easily comparable between countries, they can provide a level of detail that is not always possible with larger-scale studies. However, the national- and sub-national scale literature included in this project does not represent a comprehensive coverage of regional-based research and cannot, and should not, replace individual, detailed impacts studies in countries. The review aims to present an up-to-date assessment of the impact of climate change on each of the sectors considered.

AVOID programme results

Much of the work in this report is drawn from modelling results and analyses coming out of the AVOID programme. The AVOID programme is a research consortium funded by DECC and Defra and led by the UK Met Office and also comprises the Walker Institute at the

University of Reading, the Tyndall Centre represented through the University of East Anglia, and the Grantham Institute for Climate Change at Imperial College. The expertise in the AVOID programme includes climate change research and modelling, climate change impacts in natural and human systems, socio-economic sciences, mitigation and technology. The unique expertise of the programme is in bringing these research areas together to produce integrated and policy-relevant results. The experts who work within the programme were also well suited to review the literature assessment part of this report. In this report the modelling of sea level rise impacts was carried out for the AVOID programme by the University of Southampton.

The AVOID programme uses the same emissions scenarios across the different impact sectors studied. These are a business as usual (IPCC SRES A1B) and an aggressive mitigation (the AVOID A1B-2016-5-L) scenario. Model output for both scenarios was taken from more than 20 GCMs and averaged for use in the impact models. The impact models are sector specific, and frequently employ further analytical techniques such as pattern scaling and downscaling in the crop yield models.

Data and analysis from AVOID programme research is provided for the following impact sectors:

- Crop yields
- Water stress and drought
- Fluvial flooding
- Coastal regions

Uncertainty in climate change impact assessment

There are many uncertainties in future projections of climate change and its impacts. Several of these are well-recognised, but some are not. One category of uncertainty arises because we don't yet know how mankind will alter the climate in the future. For instance, uncertainties in future greenhouse gas emissions depends on the future socio-economic pathway, which, in turn, depends on factors such as population, economic growth, technology development, energy demand and methods of supply, and land use. The usual approach to dealing with this is to consider a range of possible future scenarios.

Another category of uncertainties relate to our incomplete understanding of the climate system, or an inability to adequately model some aspects of the system. This includes:

- Uncertainties in translating emissions of greenhouse gases into atmospheric concentrations and radiative forcing. Atmospheric CO₂ concentrations are currently rising at approximately 50% of the rate of anthropogenic emissions, with the remaining 50% being offset by a net uptake of CO₂ into the oceans and land biosphere. However, this rate of uptake itself probably depends on climate, and evidence suggests it may weaken under a warming climate, causing more CO₂ to remain in the atmosphere, warming climate further. The extent of this feedback is highly uncertain, but it not considered in most studies. The phase 3 of the Coupled Model Intercomparison Project (CMIP3), which provided the future climate projections for the IPCC Fourth Assessment Report (AR4), used a single estimate of CO₂ concentration rise for each emissions scenario, so the CMIP3 projections (which were used in most studies presented here, including AVOID) do not account for this uncertainty.
- Uncertainty in climate response to the forcing by greenhouse gases and aerosols. One aspect of this is the response of global mean temperature ("climate sensitivity"), but a more relevant aspect for impacts studies is the response of regional climates, including temperature, precipitation and other meteorological variables. Different climate models can give very different results in some regions, while giving similar results in other regions. Confidence in regional projections requires more than just agreement between models: physical understanding of the relevant atmospheric, ocean and land surface processes is also important, to establish whether the models are likely to be realistic.
- Additional forcings of regional climate. Greenhouse gas changes are not the only anthropogenic driver of climate change; atmospheric aerosols and land cover change are also important, and unlike greenhouse gases, the strength of their influence varies significantly from place to place. The CMIP3 models used in most impacts studies generally account for aerosols but not land cover change.
- Uncertainty in impacts processes. The consequences of a given changes in weather or climatic conditions for biophysical impacts such as river flows, drought, flooding, crop yield or ecosystem distribution and functioning depend on many other processes which are often poorly-understood, especially at large scales. In particular, the extent to which different biophysical impacts interact with each other has been hardly studied, but may be crucial; for example, impacts of climate change on crop yield may depend not only on local climate changes affecting rain-fed crops, but also remote climate changes affecting river flows providing water for irrigation.

Uncertainties in non-climate effects of some greenhouse gases. As well as being a greenhouse gas, CO₂ exerts physiological influences on plants, affecting photosynthesis and transpiration. Under higher CO₂ concentrations, and with no other limiting factors, photosynthesis can increase, while the requirements of water for transpiration can decrease. However, while this has been extensively studied under experimental conditions, including in some cases in the free atmosphere, the extent to which the ongoing rise in ambient CO₂ affects crop yields and natural vegetation functioning remains uncertain and controversial. Many impacts projections assume CO₂ physiological effects to be significant, while others assume it to be non-existent. Studies of climate change impacts on crops and ecosystems should therefore be examined with care to establish which assumptions have been made.

In addition to these uncertainties, the climate varies significantly through natural processes from year-to-year and also decade-to-decade, and this variability can be significant in comparison to anthropogenic forcings on shorter timescales (the next few decades) particularly at regional scales. Whilst we can characterise the natural variability it will not be possible to give a precise forecast for a particular year decades into the future.

A further category of uncertainty in projections arises as a result of using different methods to correct for uncertainties and limitations in climate models. Despite being painstakingly developed in order to represent current climate as closely as possible, current climate models are nevertheless subject to systematic errors such as simulating too little or too much rainfall in some regions. In order to reduce the impact of these, '*bias correction*' techniques are often employed, in which the climate model is a source of information on the *change* in climate which is then applied to the observed present-day climate state (rather than using the model's own simulation of the present-day state). However, these bias-corrections typically introduce their own uncertainties and errors, and can lead to inconsistencies between the projected impacts and the driving climate change (such as river flows changing by an amount which is not matched by the original change in precipitation). Currently, this source of uncertainty is rarely considered

When climate change projections from climate models are applied to climate change impact models (e.g. a global hydrological model), the climate model structural uncertainty carries through to the impact estimates. Additional uncertainties include changes in future emissions and population, as well as parameterisations within the impact models (this is rarely considered). Figure 1 highlights the importance of considering climate model structural uncertainty in climate change impacts assessment. Figure 1 shows that for 2°C prescribed global-mean warming, the magnitude of, and sign of change in average annual runoff from present, simulated by an impacts model, can differ depending upon the GCM that provides the climate change projections that drive the impact model. This example also shows that the choice of impact model, in this case a global hydrological model (GHM) or catchment-scale hydrological model (CHM), can affect the magnitude of impact and sign of change from present (e.g. see IPSL CM4 and MPI ECHAM5 simulations for the Xiangxi). To this end, throughout this review, the number of climate models applied in each study reviewed, and the other sources of uncertainty (e.g. emissions scenarios) are noted. Very few studies consider the application of multiple impacts models and it is recommended that future studies address this.

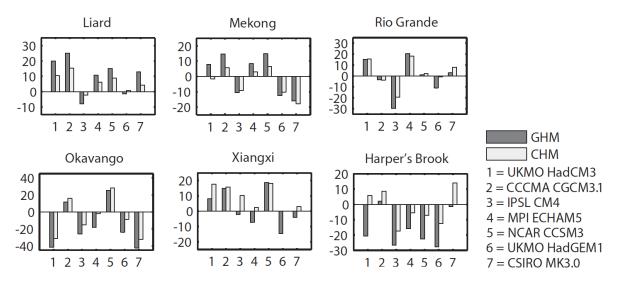


Figure 1. Change in average annual runoff relative to present (vertical axis; %), when a global hydrological model (GHM) and a catchment-scale hydrological model (CHM) are driven with climate change projections from 7 GCMs (horizontal axis), under a 2°C prescribed global-mean warming scenario, for six river catchments. The figure is from Gosling et al. (2011).

Uncertainties in the large scale climate relevant to Germany include the Atlantic Ocean has a Meridional Overturning Circulation (MOC) which transports large amounts of heat northwards in the Atlantic from the Equator. A key part of this is called the thermohaline circulation (THC). Disruption of the MOC could have a major impact on the Northern Hemisphere climate, including that of Germany, with likely detrimental impacts on human and animal systems. The IPCC AR4 concluded that "... it is very likely that the Atlantic Ocean Meridional Overturning Circulation could slow down during the course of the 21st century. A multi-model ensemble shows an average reduction of 25% with a broad range from virtually no change to a reduction of over 50% averaged over 2080 to 2099" (IPCC, 2007b).

Schneider et al. (2007) analysed simulations from several GCMs that were reviewed in the IPCC AR4 and found that projections of MOC change indicate it may weaken by 25-30% by the year 2100. Recent monitoring (Cunningham et al., 2007, Kanzow et al., 2007) has revealed large variability in the strength of the MOC on daily to seasonal timescales. This significant variability casts doubt on a previous report of decreases in MOC transport from several hydrographic sections (Bryden et al., 2005), although it does not explain the observed water mass changes below 3000m. Recent results based on radar altimeter and Argo data also suggest that there has been no slowdown, at least over the altimeter era (1993-present) (Willis, 2010). In contrast, two ocean state estimation studies (Balmaseda et al., 2007, Wunsch and Heimbach, 2006) indicated an MOC slow down. It has been suggested, based on model studies, that anthropogenic aerosols have slowed the weakening of the MOC and such weakening could only become significant several decades into the 21st century (Delworth and Dixon, 2006).

Regarding the possibility of MOC shutdown, a recent study presented by Swingedouw et al. (2007) with one climate model found that additional melt from Greenland could lead to complete AMOC shutdown in a CO₂ stabilisation experiment. However, a previous study with a different model (Ridley et al., 2005) found no effect from similar levels of meltwater input. Mikolajewicz et al. (2007) coupled an earth system model with atmospheric and ocean GCMs and observed a complete shutdown of the AMOC under a high emission scenario (SRES A2), but not before 2100. Moreover, Mikolajewicz et al. (2007) observed only a temporary weakening of the deep water formation in the North Atlantic by 2100 under a low emission scenario (B1).

Reversibility following AMOC shutdown is a key issue. Hofmann and Rahmstorf (2009) showed that hysteresis still occurs in a new low-diffusivity model. This is contrary to previous theoretical arguments that hysteresis is a product of diffusivity of the low-resolution simplified ocean models which are applied to perform the long-term simulations that are required to investigate this issue.

There is some new work on the impacts of AMOC weakening. Two studies (Kuhlbrodt et al., 2009, Vellinga and Wood, 2008) found SLR of several tens of cm along parts of the North Atlantic coast. They studies found that regional cooling could partially offset the greenhouse gas warming, and various other impacts may be substantial but hard to quantify such as

change in tropical precipitation patterns and change in ocean currents leading to declining fish stocks and ecosystems (Schmittner, 2005).

In conclusion, large uncertainty remains in the probability of a complete MOC shutdown (Kriegler et al., 2009, Zickfeld et al., 2007). However, for the high temperature scenario considered by a recent expert elicitation exercise (centred on 4.5°C by 2100, 6.5°C by 2200) (Kriegler et al., 2009), the probability of complete shutdown was assessed to be at least 10% (according to several experts). Comparable results were found by the exercise reported by Zickfeld et al. (2007). To this end, it is thought unlikely that the AMOC could significantly weaken with 2°C global-mean warming.

Summary of findings for each sector

Crop yields

- Quantitative crop yield projections for Germany under climate change scenarios vary across studies due to the application of different models, assumptions and emissions scenarios.
- Global-scale studies included here generally project yield increases by the end of the century for wheat and maize, two of Germany's major crops, and increased land suitability, with climate change. However the balance between detrimental ozone effects and CO₂ fertilisation may determine whether crop yield losses or gains are realised under climate change.
- A number of sub-national-scale assessments included here suggest that climate change could be beneficial for the wine and fruits industry in Germany.
- Important knowledge gaps and key uncertainties include the quantification of yield increases due to CO₂ fertilisation, quantification of yield reductions due to ozone damage and the extent to which crop diseases could affect crop yields with climate change.

Food security

- Germany is currently a country with extremely low levels of undernourishment. Global-scale studies included here generally project that Germany will not face serious food security issues over the next 40 years as a consequence of climate change, especially considering its strong purchasing power and substantial adaptive capacity.
- One study concluded that the national economy of Germany presents a very low vulnerability to climate change impacts on fisheries by the 2050s.

Water stress and drought

- Over 70% of Germany's current water supply is from groundwater. Global-scale studies included here which consider future groundwater withdrawals with climate change suggest that Germany is not particularly vulnerable to increased water stress in the future.
- In contrast, simulations from the AVOID programme project a median increase of around 40% of Germany's population to be exposed to water stress increases by 2100 under SRES A1B. However, the methodology for this study only accounts for changes in run-off, rather than changes in groundwater directly.
- National-scale studies included here suggest that it is possible that droughts could become more frequent with climate change in the east of the country and although annual precipitation may not change, there could be higher drought risk in summer months.

Pluvial flooding and rainfall

- Large-scale modelling studies confirm conclusions from the IPCC AR4 that heavy winter precipitation and flooding could increase with climate change for Germany.
- However, recent national-scale assessments suggest that increasing trends in precipitation with climate change for Germany are relatively small when compared with emissions uncertainty and that trends of increasing rainfall with climate change may not be significant.

Fluvial flooding

• Climate change could be associated with increases in fluvial flooding for some parts of Germany, particularly the Rhine basin.

- However, projections of future changes in flood hazard are subject to large uncertainties due to large natural variability as well as large uncertainties in the simulated climate signal. Model simulations have shown diverging responses in extreme river discharge across Germany, mostly associated with the choice of forcing climate model.
- Simulations from the AVOID programme show a greater tendency towards decreasing flood risk in Germany at first, but later in the century the models become more evenly balanced towards increasing flood risk in the A1B scenario, suggesting large uncertainty.

Tropical cyclones

• Germany is not impacted by tropical cyclones.

Coastal regions

- Germany has coasts on both the North and Baltic Seas extending over five coastal states.
- Large-scale and national-scale studies suggest that without adaptation, the number of people at risk from coastal flooding in Germany due to Sea Level Rise (SLR) under climate change scenarios could be around 300,000. However, these impacts could be reduced substantially by adaptation and climate change mitigation measures.
- One study demonstrates that while Germany is potentially highly threatened by SLR, adaptation in the form of raising of flood dykes and the application of beach nourishment could greatly reduce these impacts to levels which appear manageable.
- These studies add further detail to knowledge documented in the IPCC AR4.

Crop yields

Headline

Crop yield projections for Germany under climate change scenarios vary across studies due to the application of different models, assumptions, emissions scenarios, and crop types. However, generally, studies suggest overall yield gains and increased land crop suitability with climate change, which suggests low vulnerability. A number of sub-national scale assessments suggest that climate change could be beneficial for the wine and fruits industry in Germany.

Results from the AVOID programme for Germany indicate the balance is much more towards areas of improved rather than declining cropland suitability due to climate change, especially at 2030. The balance shifts later in the century under A1B, partly due to increased uncertainty in the areas of improved suitability and partly due to possible increased areas of declining suitability. Nevertheless the balance remains more towards improvements than declines. The smaller disagreements between models for the mitigation scenario suggest that the initial beneficial effects of low-level climate change may be more likely to remain dominant if the climate change remains low.

It should be noted, however, that the approach applied here does not consider policy and international markets. In some cases, it could be possible that what may be in extremis marginal land could still be put to use and suitable if the need for food becomes high in the future

Supporting literature

Introduction

The impacts of climate change on crop productivity are highly uncertain due to the complexity of the processes involved. Most current studies are limited in their ability to capture the uncertainty in regional climate projections, and often omit potentially important aspects such as extreme events and changes in pests and diseases. Importantly, there is a lack of clarity on how climate change impacts on drought are best quantified from an

agricultural perspective, with different metrics giving very different impressions of future risk. The dependence of some regional agriculture on remote rainfall, snowmelt and glaciers adds to the complexity - these factors are rarely taken into account, and most studies focus solely on the impacts of local climate change on rain-fed agriculture. However, irrigated agricultural land produces approximately 40-45 % of the world's food (Doll and Siebert 2002), and the water for irrigation is often extracted from rivers which can depend on climatic conditions far from the point of extraction. Hence, impacts of climate change on crop productivity often need to take account of remote as well as local climate changes. Indirect impacts via sealevel rise, storms and diseases have also not been quantified. Perhaps most seriously, there is high uncertainty in the extent to which the direct effects of CO₂ rise on plant physiology will interact with climate change in affecting productivity. Therefore, at present, the aggregate impacts of climate change on large-scale agricultural productivity cannot be reliably quantified (Gornall et al, 2010). This section summarises findings from a range of post IPCC AR4 assessments to inform and contextualise the analysis performed by AVOID programme for this project. The results from the AVOID work are discussed in the next section.

The FAO (2008) showed that wheat, barley, rapeseed, sugar beets and potatoes are Germany's most important food crops (see Table 1).

Harvested area (ha)		Quantity (Metric ton)		Value (\$1000)	
Wheat	3210000	Wheat	25900000	Wheat	2310000
Barley	1960000	Sugar beet	23000000	Potatoes	1530000
Rapeseed	1370000	Barley	11900000	Rapeseed	1430000
Rye	736000	Potatoes	11300000	Sugar beet	1050000
Maize	520000	Rapeseed	5150000	Grapes	662000
Triticale	398000	Maize	5100000	Barley	555000
Sugar beet	369000	Rye	3740000	Apples	300000

Table 1. The top 7 crops by harvested area, quantity and value according to the FAO (2008) in Germany. Crops that feature in all lists are shaded green; crops that feature in two top 7 lists are shaded amber. Data is from FAO (2008) and has been rounded down to three significant figures.

A number of impact model studies looking at crop yield which include results for some of the main crops in Germany have been conducted. They apply a variety of methodological approaches, including using different climate model inputs and treatment of other factors that might affect yield, such as impact of increased CO₂ in the atmosphere on plant growth and adaption of agricultural practises to changing climate conditions. Some studies report projections for geographic or climatic areas larger than Germany alone and it is not always clear to what extent the crop yield projections are representative for Germany only in these cases. These different models, assumptions and emissions scenarios mean that there are a range of crop yield projections for Germany.

Important knowledge gaps, which are applicable to Germany as well as at the global-scale, include; the quantification of yield increases due to CO_2 fertilisation and yield reductions due to ozone damage (Ainsworth and McGrath, 2010, Iglesias et al., 2009), the extent crop diseases could affect crop yields with climate change (Luck et al., 2011), and understanding the indirect impacts of changes in weather extremes through global market prices for agricultural goods. Most crop simulation models do not include the direct effect of extreme temperatures on crop development and growth, thus only changes in mean climate conditions are considered to affect crop yields for the studies included here.

Assessments that include a global or regional perspective

Recent past

Crop yield changes could be due to a variety of factors, which might include, but not be confined to, a changing climate. In order to assess the impact of recent climate change (1980-2008) on wheat, maize, rice and soybean, Lobell et al. (2011) looked at how the overall yield trend in these crops changed in response to changes in climate over the period studied. The study was conducted at the global-scale but national estimates for Germany were also calculated. Lobell et all. (2011) divided the climate-induced yield trend by the overall yield trend for 1980–2008, to produce a simple metric of the importance of climate relative to all other factors. The ratio produced indicates the influence of climate on the productivity trend. So for example a value of -0.1 represents a 10% reduction in yield gain due to climate change, compared to the increase that could have been achieved without climate change, but with technology and other gains. This can also be expressed as 10 years of climate trend being equivalent to the loss of roughly 1 year of technology gains. For Germany, a negative impact on wheat yield was observed relative to what would have been achieved without the climate trends (see Table 2).

Crop	Trend
Maize	0.0 to 0.1
Rice	n/a
Wheat	-0.2 to -0.1
Soybean	n/a

Table 2. The estimated net impact of climate trends for 1980-2008 on crop yields in Germany. Climate-induced yield trend divided by overall yield trend. 'n/a' infers zero or insignificant crop production or unavailability of data. Data is from Lobell et al. (2011).

Climate change studies

Several recent studies have applied climate projections from Global Climate Models (GCMs) to crop yield models to assess the global-scale impact of climate change on crop yields. Some of these studies (Iglesias and Rosenzweig, 2009, Moriondo et al., 2010) include impact estimates at the national-scale for Germany which are presented in this section. The process of CO_2 fertilisation of some crops is usually included in most climate impact studies of yields. However, other gases can influence crop yield and are not always included in impacts models. An example of this is ozone (O_3) and so a study which attempts to quantify the potential impact on crop yield of changes in ozone in the atmosphere is also included (Avnery et al., 2011). In addition to these studies, the AVOID programme analysed the patterns of climate change for 21 GCMs, to establish an index of 'climate suitability' of agricultural land. Climate suitability is not directly equivalent to crop yields, but is a means of looking at a standard metric across all the countries including in this project, and of assessing the level of agreement on variables that affect crop production, between all 21 GCMs.

Iglesias and Rosenzweig (2009) repeated an earlier study presented by Parry et al. (2004) by applying climate projections from the HadCM3 GCM (instead of HadCM2, which was applied by Parry et al. (2004)), under seven SRES emissions scenarios and for three future time periods. This study used a globally consistent crop simulation methodologies and climate change scenarios, and weighted the model site results by their contribution to regional and national, and rain-fed and irrigated production. The study also applied a quantitative estimation of physiological CO₂ effects on crop yields and considered the affect of adaptation by assessing the country or regional potential for reaching optimal crop yield. The results from the study for Germany are presented in Table 3 and Table 4. Wheat yield was projected to increase above baseline (1970-2000) levels for each future time horizon and each scenario, with the highest yield gains in the A2 scenarios. Maize yields in 2020 and 2050 were slightly (<5%) lower than baseline level but by 2080 a small gain was estimated with the A1FI and A2 scenarios, but not with the B1 and B2 scenarios.

Scenario	Year	Wheat	Maize
	2020	4.19	-1.68
	2050	9.28	-1.37
A1FI	2080	7.48	1.19
	2020	5.67	-1.16
	2050	9.20	-1.32
A2a	2080	13.14	0.45
	2020	3.49	-0.72
	2050	8.92	-1.66
A2b	2080	13.15	0.77
	2020	3.34	-1.48
	2050	9.07	-1.81
A2c	2080	13.51	0.44
	2020	1.61	-1.71
	2050	5.28	-2.83
B1a	2080	6.86	-3.21
	2020	3.66	-2.60
	2050	5.13	-3.19
B2a	2080	7.24	-1.42
	2020	3.16	-2.55
	2050	5.50	-3.13
B2b	2080	8.85	-1.35

Table 3. Wheat and maize yield changes (%) in Germany relative to baseline scenario (1970-2000) for different emission scenarios and future time periods, for Germany. Some emissions scenarios were run in an ensemble simulation (e.g. A2a, A2b, A2c). Data is from Iglesias and Rosenzweig (2009).

	Wheat Up Down		Maize	
			Up	Down
Baseline to 2020	7	0	0	7
Baseline to 2050	7	0	0	7
Baseline to 2080	7	0	4	3
2020 to 2050	7	0	1	6
2050 to 2080	6	1	6	1

Table 4. The number of emission scenarios that predict yield gains ("Up") or yield losses ("Down") for wheat and maize in Germany between two points in time. Data is from Iglesias and Rosenzweig (2009).

Moriondo et al. (2010) simulated relative changes in crop yield for sunflower, soybean, spring wheat and durum wheat for a global mean warming of 2°C warmer than present climate change scenario with A2 socioeconomics. The study accounted for changes in extreme events such as droughts and the CO_2 fertiliser effect. Moriondo et al. (2010) compared the effectiveness of various adaptation options relative to no adaptation. No quantitative information on impacts is available from the study but estimates can be made whether, on average, a relative yield loss or a yield gain was projected for a given crop,

adaptation method, and country (see Table 5). For Germany, the results indicate that for the 2030-2060 time horizon and without adaptation, yield losses for soybean and potentially for sunflower and spring wheat are possible with climate change. However, advanced sowing, longer cycle varieties and irrigation could offset these yield losses for sunflower and spring wheat. For soybean, longer cycle varieties and irrigation appeared to have potential to offset crop yield losses but this cannot be estimated quantitatively from the study. It should be noted, however, that only a small part of Germany is included in the analysis presented by Moriondo et al. (2010), so the results should not be considered as representative for the whole of Germany.

	No adaptation ¹	Advanced sowing	Delayed sowing	Shorter cycle varieties	Longer cycle varieties	Irrigation
Sunflower	+ -	+	-	-	+	+
Soybean	-	+	-	-	+	+
Spring wheat	+ -	+	-	-	+	+
Durum wheat	n/a	n/a	n/a	n/a	n/a	n/a

¹ Yield changes with respect to the present period, not considering adaptation methods

Table 5. Relative change in yield of four crops in a +2 °C world under SRES A2 socioeconomics for Germany. The relative change is calculated with respect to the same +2°C scenario without adaptation (left column). "+" = relative yield gain, "-" = relative yield loss, "+ -" = high spatial variability and uncertainty over sign of average yield change, "n/a" = crop is not grown. After Moriondo et al. (2010).

Elsewhere, several recent studies have assessed the impact of climate change on a globalscale or regional-scale and include impact estimates for Western Europe as a whole. Whilst these studies provide a useful indicator of crop yields under climate change for the larger *region*, it should be noted that the crop yields presented in such cases are not definitive *national* estimates. This is because the yields are averaged over the entire region, which includes other countries as well as Germany.

Tatsumi et al. (2011) applied an improved version of the GAEZ crop model (iGAEZ) to simulate crop yields on a global scale for wheat, potato, cassava, soybean, rice, sweet potato, maize, green beans. The impact of global warming on crop yields from the 1990s to 2090s was assessed by projecting five GCM outputs under the SRES A1B scenario and comparing the results for crop yields as calculated using the iGAEZ model for the period of 1990-1999. The results for Western Europe, the regional grouping which includes Germany,

are displayed in Table 6 and suggest increased yields for wheat, potato and green beans, but a decline for rice and maize.

Wheat	Potato	Cassava	Soybean	Rice	Sweet potato	Maize	Green beans
7.81	4.07	-	-0.03	-1.11	-	-5.17	7.34

Table 6. Average change in yield (%), during 1990s-2090s in Western Europe. Data is from Tatsumi et al. (2011).

Olesen et al (2007) addressed the issue of uncertainty in projecting impacts of climate change on agriculture. They projected rain-fed winter wheat yield across the European domain using nine different RCMs with HadAM3H as the bounding GCM, under SRES A2 emissions. For more than 90% of the cropping area of Germany all RCMs projected an increase in wheat yield.

The PESETA project estimated the impacts of climate change on crop yields for different regions in the EU (Ciscar et al., 2009, Iglesias et al., 2009). Climate scenarios were created for the 2070-2100 time horizon using a combination of two GCMs and SRES emissions scenarios (A2 and B2). Crop yield simulations (winter wheat, spring wheat, rice, grassland, maize and soybeans) were then conducted using the DSSAT suite of crop models. The results for "Central Europe North", which includes Germany and other countries, are displayed in Table 7. As mentioned previously, it should be noted that the projected yield changes may vary widely within a geographic region. The Central Europe North average is probably not exactly representative for Germany alone. Nevertheless, the PESETA project includes useful maps that show projected changes in crop yield for each emissions scenario, from which impacts for Germany can be inferred (see Figure 2). The results suggest that, dependent on region and scenario climate change may have either a small beneficial impact or a small negative effect on crop yields in Germany.

2011-2040	2071-2100				
A2 ECHAM4	A2 HadAM3h	B2 HadAM3h	A2 ECHAM4	B2 ECHAM4	
+16	-3	-1	-8	+2	

Table 7. Projected crop yield changes (%), compared to 1961-1990 period for the "Central Europe North" region, which includes Germany. Data is from Ciscar et al. (2009).

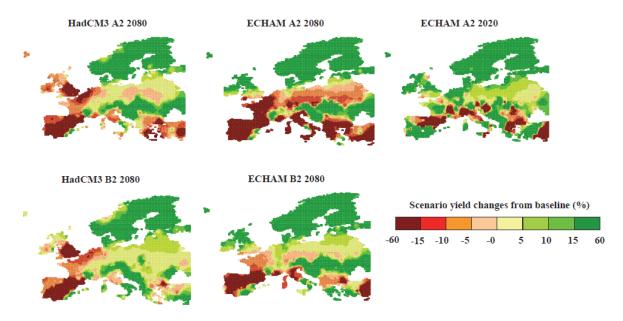


Figure 2. Crop yield changes under the HadCM3/HIRHAM A2 and B2 scenarios for the period_2071 - 2100 and for the ECHAM4/RCA3 A2 and B2 scenarios for the period 2011 – 2040_compared with the baseline. The figure is from Iglesias et al., 2009, p.31.

In addition to the studies looking at the effect of changes in climate and CO_2 concentrations on crop yield, Avnery et al. (2011) investigated the effects of ozone surface exposure on crop yield losses for soybeans, maize and wheat under the SRES A2 and B1 scenarios respectively. Two metrics of ozone exposure were investigated; seasonal daytime (08:00-19:59) mean O_3 ("M12") and accumulated O_3 above a threshold of 40 ppbv ("AOT40"). The effect of the ozone exposure was considered in isolation from climate and other changes. The results for Germany are presented in Table 8.

	A	2	B1		
	M12 AOT40		M12	AOT40	
Soybeans	-	-	-	-	
Maize	6-8	4-6	6-8	2-4	
Wheat	4-6	15-20	2-4	10-15	

Table 8. National relative crop yield losses (%) for 2030 under A2 and B1 emission scenarios according to the M12 (seasonal daytime (08:00–19:59) mean) and AOT40 (accumulated O_3 above a threshold of 40 ppbv) metrics of O_3 exposure. Data is from Avnery et al. (2011).

National-scale or sub-national scale assessments

Climate change studies

In this section we present results from recent studies that have produced national or sub-

national scale projections of future crop yields in Germany.

Research suggests that the vineyards of the Mosel Valley could increase in value under a scenario of global warming, and perhaps by a considerable amount. showed that a moderate 1°C temperature increase could lead to an aggregate increase in land value of 20% or more, while an increase of 3°C could more than double the land value in the region. The increases in value were due to an increase in wine quality, associated with changes in temperature and solar radiation. This supports the general contention that crop suitability could increase in Germany with climate change, proposed by results from the AVOID programme (see Figure 3).

However, there could be regional differences in the response of German crop yields to climate change, even at the sub-national level. For example, Döll and Schulze (2010) discuss how even for the Hamburg metropolitan region, climate change may have differential impacts, depending upon water availability and the CO₂ fertilisation effect. The authors showed that fruit yields in the region may increase with climate change, which broadly supports results in a study presented by Kemfert et al. (2009), that showed apple production in the Alte Land could benefit from climate change. Kemfert et al. (2009) found that increases in minimum regional temperature during the annual blossoming season, and increases in mean regional precipitation could lead to higher land productivity levels.

Moreover, Lang (2007) notes that German farmers may benefit from climate change in the short-term, with maximum gains occurring at a temperature increase of 0.6°C against current levels. In the longer-term, there may be losses from global warming, but Lang (2007) concludes that the net present value from climate change is, under the most likely future scenarios, positive. Likewise, simulations presented by Wechsung et al. (2008) show that both corn and wheat yields in eastern Germany in the medium term, i.e. in the next 20-30 years, could remain stable with climate change. However, the authors note that by the middle of the century, locations in the eastern lowlands may experience declines in yield. Similarly, Kersebaum et al. (2009) showed that under climate conditions simulated by the WETTREG climate model, under A1B emissions, and with CO₂ fertilisation, wheat yields could increase in many regions across Germany. However, when the effect of CO₂ fertilisation was omitted, crop yields declined relative to present yields. Similarly, research from the REGKLAM project ("Development and Testing of an Integrated Regional Climate Change Adaptation Programme for the Model Region of Dresden", in German) showed that projections of crop yields for Saxony up to 2050, were affected significantly less by climate

change projections than by assumptions on the CO₂ fertilisation effect or future technological progress (Köstner et al., 2010).

AVOID programme results

To further quantify the impact of climate change on crops, the AVOID programme simulated the effect of climate change on the suitability of land for crop cultivation for all countries reviewed in this literature assessment based upon the patterns of climate change from 21 GCMs (Warren et al., 2010). This ensures a consistent methodological approach across all countries and takes consideration of climate modelling uncertainties.

Methodology

The effect of climate change on the suitability of land for crop cultivation is characterised here by an index which defines the percentage of cropland in a region with 1) a decrease in suitability or 2) an increase in suitability. A threshold change of 5% is applied here to characterise decrease or increase in suitability. The crop suitability index is calculated at a spatial resolution of 0.5°x0.5°, and is based on climate and soil properties (Ramankutty et al., 2002). The baseline crop suitability index, against which the future changes are measured, is representative of conditions circa 2000. The key features of the climate for the crop suitability index are temperature and the availability of water for plants. Changes in these were derived from climate model projections of future changes in temperature and precipitation, with some further calculations then being used to estimate actual and potential evapotranspiration as an indicator of water availability. It should be noted that changes in atmospheric CO₂ concentrations can decrease evapotranspiration by increasing the efficiency of water use by plants (Ramankutty et al., 2002), but that aspect of the index was not included in the analysis here. Increased CO₂ can also increase photosynthesis and improve yield to a small extent, but again these effects are not included. Exclusion of these effects may lead to an overestimate of decreases in suitability.

The index here is calculated only for grid cells which contain cropland circa 2000, as defined in the global crop extent data set described by Ramankutty et al. (2008) which was derived from satellite measurements. It is assumed that crop extent does not change over time. The crop suitability index varies significantly for current croplands across the world (Ramankutty et al., 2002), with the suitability being low in some current cropland areas according to this index. Therefore, while climate change clearly has the potential to decrease suitability for cultivation if temperature and precipitation regimes become less favourable, there is also scope for climate change to increase suitability in some existing cropland areas if conditions become more favourable in areas where the suitability index is not at its maximum value of 1. It should be noted that some areas which are not currently croplands may already be suitable for cultivation or may become suitable as a result of future climate change, and may become used a croplands in the future either as part of climate change adaptation or changes in land use arising for other reasons. Such areas are not included in this analysis.

Results

Crop suitability was estimated under the pattern of climate change from 21 GCMs with two emissions scenarios; 1) SRES A1B and 2) an aggressive mitigation scenario where emissions follow A1B up to 2016 but then decline at a rate of 5% per year thereafter to a low emissions floor (denoted A1B-2016-5-L). The application of 21 GCMs is an attempt to quantify the uncertainty due to climate modelling, although it is acknowledged that only one crop suitability impacts model is applied. Simulations were performed for the years 2030, 2050, 2080 and 2100. The results for Germany are presented in Figure 3.

Under all the climate projections, some existing cropland areas in Germany become more suitable for cultivation, while a few models show some of the existing cropland areas become less suitable, mainly under the A1B emission scenario. The areas of increased and decreased suitability differ considerably according to the climate model used.

In 2030, under both the A1B and aggressive mitigation scenarios, the model projections imply an improvement in suitability for cultivation between 50%-70% of current German croplands. Under the aggressive mitigation scenario, this increases to 60% to 75% by the end of the century. However, under the A1B scenario, the difference between models becomes much larger over time. The mean projected area of improved suitability for A1B increases from 63% in 2030 to 71% in 2100. The mitigation scenario generally implies no areas with declining suitability (only 1 model implies declining suitability and then only over 2% of current German croplands by 2100). Under the A1B scenario only, no areas of declining suitability are projected for 2030 but up to 27% of current croplands are projected to become less suitable by 2100.

So, for Germany, the balance is much more towards areas of improved rather than declining cropland suitability due to climate change. The balance shifts later in the century under A1B, partly due to increased uncertainty in the areas of improved suitability and partly due to possible increased areas of declining suitability. Nevertheless the balance remains more

towards improvements than declines. The smaller disagreements between models for the mitigation scenario suggest that the initial beneficial effects of low-level climate change may be more likely to remain dominant if the climate change remains low.

It should be noted, however, that the approach applied here does not consider policy and international markets. In some cases, it could be possible that what may be in extremis marginal land could still be put to use and suitable if the need for food becomes high in the future.

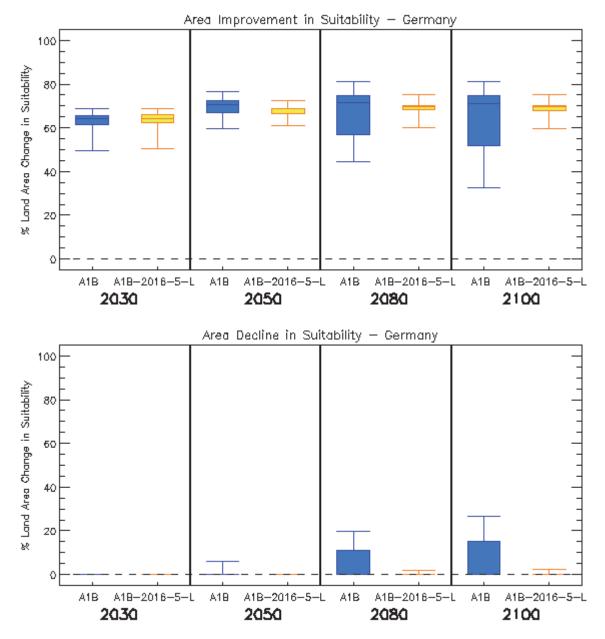


Figure 3. Box and whisker plots for the impact of climate change on increased crop suitability (top panel) and decreased crop suitability (bottom panel) for Germany, from 21 GCMs under two emissions scenarios (A1B and A1B-2016-5-L), for four time horizons. The plots show the 25th, 50th, and 75th percentiles (represented by the boxes), and the maximum and minimum values (shown by the extent of the whiskers).

Food security

Headline

Studies suggest that Germany possesses a high capability of importing food due to strong purchasing power and financial support. Also, Germany has substantial adaptive capacity and possesses proactive food management systems. These factors mean that Germany could remain food-secure with climate change.

Supporting literature

Introduction

Food security is a concept that encompasses more than just crop production, but is a complex interaction between food availability and socio-economic, policy and health factors that influence access to food, utilisation and stability of food supplies. In 1996 the World Food Summit defined food security as existing 'when all people, at all times, have physical and economic access to sufficient, safe and nutritious food to meet their dietary needs, and their food preferences are met for an active and healthy life'. As such this section cannot be a comprehensive analysis of all the factors that are important in determining food security, but does attempt to assess a selection of the available literature on how climate change, combined with projections of global and regional population and policy responses, may influence food security.

With regards to food security Germany is a country of very low concern, relative to other countries across the globe. According to FAO statistics (FAO, 2010) Germany has extremely low rates of undernourishment (less than 5% of the population). Moreover, a number of global studies point towards a generally optimistic and positive outlook for the impact of climate change on food security in Germany, largely as a result of its high adaptive capacity and its ability to be able to afford to import food to offset potential deficits in food production.

Assessments that include a global or regional perspective

Several recent studies have analysed food security under climate change across the globe. For example, Wu et al. (2011) simulated crop yields with the GIS-based Environmental Policy Integrated Climate (EPIC) model. This was combined with crop areas simulated by a crop choice decision model to calculate total food production and per capita food availability across the globe, which was used to represent the status of food availability and stability. The study focussed on the SRES A1 scenario and applied climate change simulations for the 2000s (1991–2000) and 2020s (2011–2020). The climate simulations were performed by MIROC (Model for Interdisciplinary Research on Climate) version 3.2, which means the effects of climate model uncertainty were not considered. Downscaled population and GDP data from the International Institute for Applied Systems Analysis (IIASA) were applied in the simulations. Whilst Germany appeared as a hot spot for food security vulnerability in 2020, Wu et al. (2011) note that Germany's population may still be food-secure as their population is less reliant on subsistence agriculture and because Germany possess a high capability of importing food due to strong purchasing power and financial support. Also, Germany presents substantial adaptive capacity and possesses proactive food management systems.

Falkenmark et al. (2009) present a global analysis of food security under climate change scenarios for the 2050s that considers the importance of water availability for ensuring global food security. The study presents an analysis of water constraints and opportunities for global food production on current croplands and assesses five main factors:

- how far improved land and water management might go towards achieving global food security,
- 2) the water deficits that would remain in regions currently experiencing water scarcity and which are aiming at food self-sufficiency,
- 3) how the water deficits above may be met by importing food,
- the cropland expansion required in low income countries without the needed purchasing power for such imports, and
- 5) the proportion of that expansion pressure which will remain unresolved due to potential lack of accessible land.

Similar to the study presented by Wu et al. (2011), there is no major treatment of modelling uncertainty; simulations were generated by only the LPJml dynamic global vegetation and

water balance model (Gerten et al., 2004) with population growth and climate change under the SRES A2 emission scenario. Falkenmark et al. (2009) summarise the impacts of future improvements (or lack thereof) in water productivity for each country across the globe and show that this generates either a deficit or a surplus of water in relation to food water requirements in each country. These can be met either by trade or by horizontal expansion (by converting other terrestrial ecosystems to crop land). The study estimated that in 2050 around one third of the world's population will live in each of three regions: those that export food, those that import food, and those that have to expand their croplands at the expense of other ecosystems because they do not have enough purchasing power to import their food. The simulations demonstrated that Germany was a food importing country in 2050.

The International Food Policy Research Institute (IFPRI) have produced a report and online tool that describes the possible impact of climate change on two major indicators of food security; 1) the number of children aged 0 to 5 malnourished, and 2) the average daily kilocalorie availability (Nelson et al., 2010, IFPRI, 2010). The study considered three broad socio-economic scenarios; 1) a 'pessimistic' scenario, which is representative of the lowest of the four GDP growth rate scenarios from the Millennium Ecosystem Assessment GDP scenarios and equivalent to the UN high variant of future population change, 2) a 'baseline' scenario, which is based on future GDP rates estimated by the World Bank and a population change scenario equivalent to the UN medium variant, and 3) an 'optimistic' scenario that is representative of the highest of the four GDP growth rate scenarios from the Millennium Ecosystem Assessment GDP scenarios and equivalent to the UN low variant of future population change. Nelson et al. (2010) also considered climate modelling and emission uncertainty and included a factor to account for CO₂ fertilisation in their work. The study applied two GCMs, the CSIRO GCM and the MIROC GCM, and forced each GCM with two SRES emissions scenarios (A1B and B1). They also considered a no climate change emissions scenario, which they called 'perfect mitigation' (note that in most other climate change impact studies that this is referred to as the baseline). The perfect mitigation scenario is useful to compare the effect of climate change against what might have happened without, but is not a realistic scenario itself. IFPRI have not published projections for child malnourishment in Germany but information on average daily kilocalorie availability has been made available. Table 9 displays the average daily kilocalorie availability simulated under different climate and socioeconomic scenarios for Germany and Figure 4 displays the effect of climate change, calculated by comparing the 'perfect mitigation' scenario with each baseline, optimistic and pessimistic scenario. Whilst by 2050 climate change is attributable for up to around a 7% decline in kilocalorie availability, the absolute value of available

kilocalories remains well above 3,000 under all but two scenarios. This suggests Germany may not face food security issues in 2050.

Scenario	2010	2050
Baseline CSI A1B	3289	3266
Baseline CSI B1	3294	3287
Baseline MIR A1B	3275	3200
Baseline MIR B1	3285	3250
Baseline Perfect Mitigation	3328	3453
Pessimistic CSI A1B	3331	3028
Pessimistic CSI B1	3335	3046
Pessimistic MIR A1B	3316	2968
Pessimistic MIR B1	3323	2999
Pessimistic Perfect Mitigation	3370	3192
Optimistic CSI A1B	3287	3385
Optimistic CSI B1	3290	3400
Optimistic MIR A1B	3272	3309
Optimistic MIR B1	3279	3343
Optimistic Perfect Mitigation	3325	3573

Table 9. Average daily kilocalorie availability simulated under different climate and socioeconomic scenarios, for Germany (IFPRI, 2010).

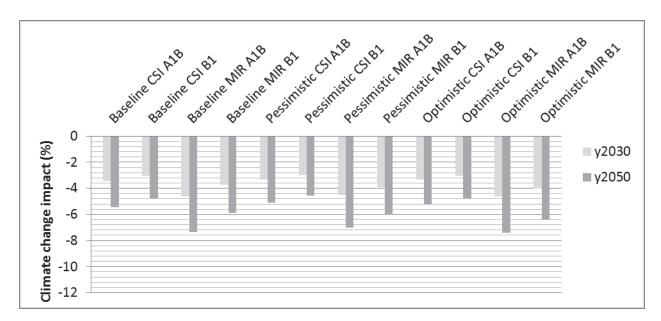


Figure 4. The impact of climate change on average daily kilocalorie availability for Germany (IFPRI, 2010).

It is important to note that up until recently, projections of climate change impacts on global food supply have tended to focus solely on production from terrestrial biomes, with the large

contribution of animal protein from marine capture fisheries often ignored. However, recent studies have addressed this knowledge gap (e.g., Allison et al., 2009). In addition to the direct affects of climate change, changes in the acidity of the oceans, due to increases in CO₂ levels, could also have an impact of marine ecosystems, which could also affect fish stocks. However, this relationship is complex and not well understood, and studies today have not been able to begin to quantify the impact of ocean acidification on fish stocks.

Allison et al. (2009) present a global analysis that compares the vulnerability of 132 national economies to potential climate change impacts on their capture fisheries. The study considered a country's vulnerability to be a function of the combined effect of projected climate change, the relative importance of fisheries to national economies and diets, and the national societal capacity to adapt to potential impacts and opportunities. Climate change projections from a single GCM under two emissions scenarios (SRES A1FI and B2) were used in the analysis. Allison et al. (2009) concluded that the national economy of Germany presented a very low vulnerability to climate change impacts on fisheries. In contrast, countries in Central and Western Africa (e.g. Malawi, Guinea, Senegal, and Uganda), Peru and Colombia in north-western South America, and four tropical Asian countries (Bangladesh, Cambodia, Pakistan, and Yemen) were identified as most vulnerable (see Figure 5). It should be noted, however, that results from studies that have applied only a single climate model or climate change scenario should be interpreted with caution. This is because they do not consider other possible climate change scenarios which could result in a different impact outcome, in terms of magnitude and in some cases sign of change.

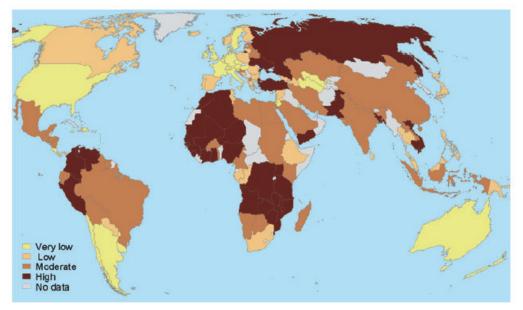


Figure 5. Vulnerability of national economies to potential climate change impacts on fisheries under SRES B2 (Allison et al., 2009). Colours represent quartiles with dark brown for the upper quartile (highest index value), yellow for the lowest quartile, and grey where no data were available.

National-scale or sub-national scale assessments

Literature searches yielded no results for national-scale or sub-national scale studies for this impact sector.

Water stress and drought

Headline

Over 70% of Germany's water supply is from groundwater. Two global-scale studies, both of which considered future groundwater withdrawals with climate change, suggest that Germany is not particularly vulnerable to increased water stress from climate change. National-scale studies suggest that it is possible that droughts could become more frequent with climate change in the east of the country and although annual precipitation may not change, there could be higher drought risk in summer months.

Results from the AVOID programme for Germany indicate a growing proportion of the population experiencing increased water stress. However, the methodology for this study only accounts for changes in runoff, rather than changes in groundwater directly, so should be viewed in this light.

Supporting literature

Introduction

For the purposes of this report droughts are considered to be extreme events at the lower bound of climate variability; episodes of prolonged absence or marked deficiency of precipitation. Water stress is considered as the situation where water stores and fluxes (e.g. groundwater and river discharge) are not replenished at a sufficient rate to adequately meet water demand and consumption.

A number of impact model studies looking at water stress and drought for the present (recent past) and future (climate change scenario) have been conducted. These studies are conducted at global or national scale and include the application of global water 'availability' or 'stress' models driven by one or more climate change scenario from one or more GCM. The approaches variously include other factors and assumptions that might affect water availability, such as the impact of changing demographics and infrastructure investment, etc. These different models (hydrological and climate), assumptions and emissions scenarios mean that there are a range of water stress projections for Germany. This section summarises findings from these studies to inform and contextualise the analysis performed by the AVOID programme for this project. The results from the AVOID work and discussed in the next section.

Important knowledge gaps and key uncertainties which are applicable to Germany as well as at the global-scale, include; the appropriate coupling of surface water and groundwater in hydrological models, including the recharge process, improved soil moisture and evaporation dynamics, inclusion of water quality, inclusion of water management (Wood et al., 2011) and further refinement of the down-scaling methodologies used for the climate driving variables (Harding et al., 2011).

Assessments that include a global or regional perspective

Recent past

Recent research presented by Vörösmarty et al. (2010) describes the calculation of an 'Adjusted Human Water Security Threat' (HWS) indicator. The indicator is a function of the cumulative impacts of 23 biophysical and chemical drivers simulated globally across 46,517 grid cells representing 99.2 million km². With a digital terrain model at its base, the calculations in each of the grid boxes of this model take account of the multiple pressures on the environment, and the way these combine with each other, as water flows in river basins. The level of investment in water infrastructure is also considered. This infrastructure measure (the *investment benefits factor*) is based on actual existing built infrastructure, rather than on the financial value of investments made in the water sector, which is a very unreliable and incomplete dataset. The analysis described by Vörösmarty et al. (2010) represents the current state-of-the-art in applied policy-focussed water resource assessment. In this measure of water security, the method reveals those areas where this is lacking, which is a representation of human water stress. One drawback of this method is that no analysis is provided in places where there is 'no appreciable flow', where rivers do not flow, or only do so for such short periods that they cannot be reliably measured. This method also does not address places where water supplies depend wholly on groundwater or desalination, being piped in, or based on wastewater reuse. It is based on what is known from all verified peer reviewed sources about surface water resources as generated by natural ecosystem processes and modified by river and other hydraulic infrastructure (Vörösmarty et al., 2010).

Here, the present day HWS for is mapped for Germany. The model applied operates at 50km resolution, so, larger countries appear to have smoother coverage than smaller countries, but all are mapped and calculated on the same scale, with the same data and model, and thus comparisons between places are legitimate. It is important to note that this analysis is a comparative one, where each place is assessed *relative* to the rest of the globe. In this way, this presents a realistic comparison of conditions across the globe. As a result of this, however, some places may seem to be less stressed than may be originally considered.

Figure 6 presents the results of this analysis for Germany. Although large areas of the country are shown to exhibit a moderate water security threat, it must be considered that this analysis does not consider groundwater resources, which Germany depends upon for 70% of its total water withdrawals (Aljazzar, 2010, Zhu and Balke, 2008).

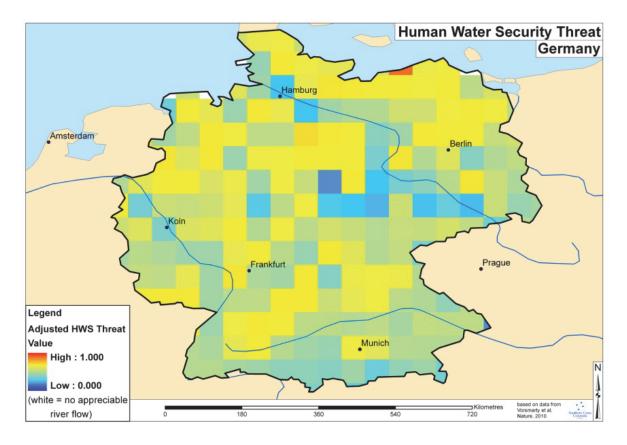


Figure 6. Present Adjusted Human Water Security Threat (HWS) for Germany, calculated following the method described by Vörösmarty et al. (2010).

Smakhtin et al. (2004) present a first attempt to estimate the volume of water required for the maintenance of freshwater-dependent ecosystems at the global scale. This total environmental water requirement (EWR) consists of ecologically relevant low-flow and high-

flow components. The authors argue that the relationship between water availability, total use and the EWR may be described by the water stress indicator (WSI). If WSI exceeds 1.0, the basin is classified as "environmentally water scarce". In such a basin, the discharge has already been reduced by total withdrawals to such levels that the amount of water left in the basin is less than EWR. Smaller index values indicate progressively lower water resources exploitation and lower risk of "environmental water scarcity." Basins where WSI is greater than 0.6 but less than 1.0 are arbitrarily defined as heavily exploited or "environmentally water stressed" and basins where WSI is greater than 0.3 but less than 0.6 are defined as moderately exploited. In these basins, 0-40% and 40-70% of the utilizable water respectively is still available before water withdrawals come in conflict with the EWR. Environmentally "safe" basins are defined as those where WSI is less than 0.3. The global distribution of WSI for the 1961-1990 time horizon is shown in Figure 7. The results show that for the basins considered, Germany exhibits moderate to high water stress, but it should be noted that this refers to maintenance of freshwater-dependent ecosystems, rather than water availability for human society.

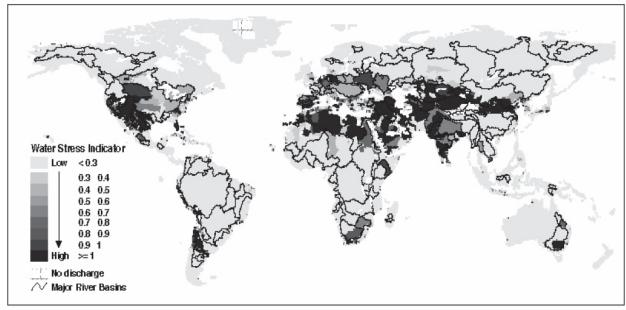


Figure 7. A map of the major river basins across the globe and the water stress indicator (WSI) for the 1961-1990 time horizon. The figure is from Smakhtin et al. (2004).

Climate change studies

Rockström et al. (2009) applied the LPJml vegetation and water balance model (Gerten et al., 2004) to assess green-blue water (irrigation and infiltrated water) availability and requirements. The authors applied observed climate data from the CRU TS2.1 gridded dataset for a present-day simulation, and climate change projections from the HadCM2 GCM

under the SRES A2 scenario to represent the climate change scenario for the year 2050. The study assumed that if water availability was less than 1,300m³/capita/year, then the country was considered to present insufficient water for food self-sufficiency. The simulations presented by Rockström et al.(2009) should not be considered as definitive, however, because the study only applied one climate model, which means climate modelling uncertainty was overlooked. The results from the two simulations are presented in Figure 8. Rockström et al. (2009) found that globally in 2050 and under the SRES A2 scenario, around 59% of the world's population would be exposed to "blue water shortage" (i.e. irrigation water shortage and shortage of water from groundwater), and 36% exposed to "green water shortages" (i.e. infiltrated rain shortage). For Germany, Rockström et al. (2009) found that blue-green water availability was around the 1,300m³/capita/year threshold in present but that with climate change, this would increase to above 1,700m³/capita/year, thereby meaning that at a national level, Germany's water resource requirements should be met by 2050.

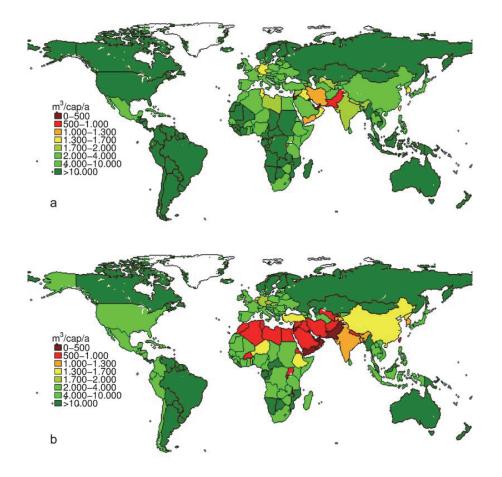


Figure 8. Simulated blue-green water availability (m^3 /capita/year) for present climate (a) and including both demographic and climate change under the SRES A2 scenario in 2050 (b). The study assumed that if water availability was less than 1,300 m^3 /capita/year, then the country was considered to present insufficient water for food self-sufficiency. The figure is from Rockström et al. (2009).

Doll (2009) presents updated estimates of the impact of climate change on groundwater resources by applying a new version of the WaterGAP hydrological model. The study accounted for the number of people affected by changes in groundwater resources under climate change relative to present (1961-1990). To this end, the study provides an assessment of the vulnerability of humans to decreases in available groundwater resources (GWR). This indicator was termed the "Vulnerability Index" (VI), defined as; VI = -% change GWR * Sensitivity Index (SI). The SI component was a function of three more specific sensitivity indicators that include an indicator of water scarcity (calculated from the ratio between consumptive water use to low flows), an indicator for the dependence upon groundwater supplies, and an indicator for the adaptive capacity of the human system. Doll (2009) applied climate projections from two GCMs (ECHAM4 and HadCM3) to WaterGAP, for two scenarios (SRES A2 and B2), for the 2050s. Figure 9 presents each of these four simulations respectively. There is variation across scenarios and GCMs. With HadCM3, the

simulated VI for Germany is negligible, because the simulated GWR decreases by less than 10% with climate change. However, it should be noted that these simulations are based upon projections from only two GCMs; projections from other GCMs could yield different results.

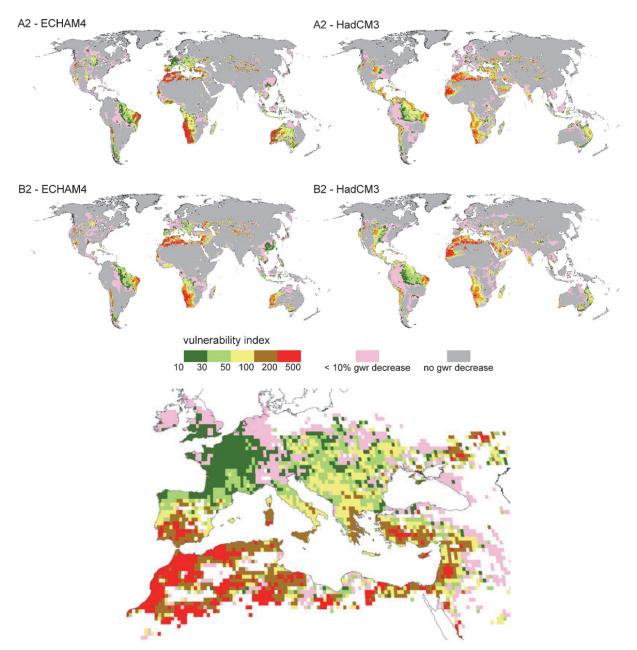


Figure 9. Vulnerability index (VI) showing human vulnerability to climate change induced decreases of renewable groundwater resources (GWR) by the 2050s under two emissions scenarios for two GCMs. VI is only defined for areas with a GWR decrease of at least 10% relative to present (1961-1990). Also shown is VI for the Mediterranean region with ECHAM4 under A2 emissions. The figure is from Doll (2009).

Lehner et al. (2006) assessed the impact of climate change on European drought risk. The authors accounted for future human water use and assessed future flood and drought frequencies by applying the WaterGAP hydrological model, driven by climate projections from the HadCM3 and ECHAM4 GCMs, under a 1%/year CO₂ increase emissions scenario. The simulations are presented in Figure 10 and Figure 11. The results reflect the general consensus from other studies that southern and south-eastern Europe could experience increased drought frequencies, leading to water stress. This in part due to increased water use but the impacts are much more pronounced and wide spread when climate change is factored in (Lehner et al., 2006). Little or no change in the intensity of 100-year droughts is projected for Germany by the 2070s compared to baseline estimates, and without climate change there is no projected change across almost the entire country. However when considering the change in recurrence of 100-year droughts to become more frequent in the east of the country.

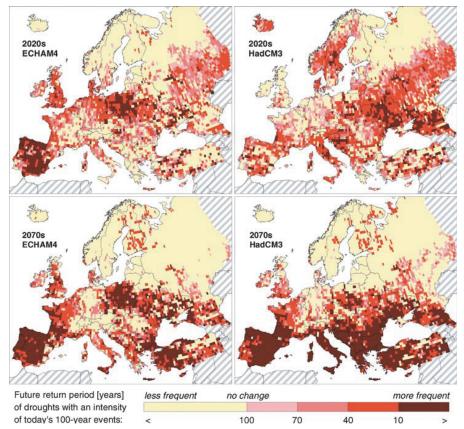


Figure 10. Change in recurrence of 100-year droughts, based on comparisons between today's climate and water use (1961–1990) and simulations for the 2020s and 2070s (ECHAM4 and HadCM3 GCMs), under a 1%/year CO₂ increase emissions scenario. The figure is from Lehner et al. (2006).

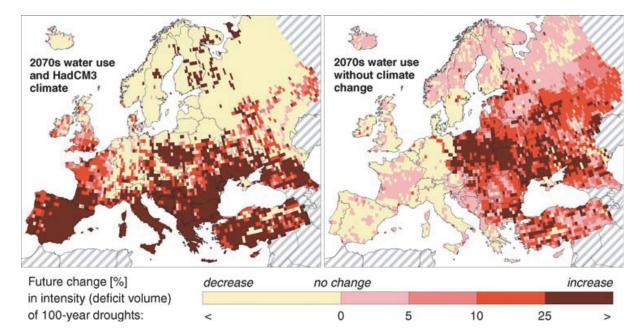


Figure 11. Change in intensity of 100-year droughts, based on comparison between today's climate and water use (1961–1990) and simulations for the 2070s (left map: HadCM3 GCM; right map: only water use scenario, no climate change), under a 1%/year CO₂ increase emissions scenario.

National-scale or sub-national scale assessments

The RheinBlick2050 project (Görgen et al., 2010) assessed the impacts of future climate change on discharge of the Rhine River and its major tributaries. The project applied an ensemble of 20 dynamically downscaled transient bias-corrected regional climate simulations (control runs and projections) as forcing data for the HBV134 hydrological model at a daily temporal resolution over the catchments of the Rhine River. The authors of the report note that discharge projections presented here give an indication of the changes in the hydro-meteorological system of the Rhine River and its catchments as a result of climate change projections that are regarded the state-of-the-art in 2010. With respect to low flow, Görgen et al. (2010) found no strong decrease by the 2030s; while most ensemble members showed no clear tendency in summer (ranging from +/-10%), winter low flow was projected to be higher (0% to +15%). For the 2080s, the change signal was stronger in summer, with a tendency towards decreased low flow discharges (-25% to 0%), while for winter no clear signal was discernible (range in projections mainly from -5% to +20% depending on discharge diagnostic and gauging station).

Similar conclusions are noted in a review by Leipprand et al. (2008). The authors found that while Germany might have to face more floods in the Rhine basin, particular in winter, reductions in average river discharges might be experienced for the summer months. In the eastern parts of the country, the Elbe basin, reductions in water availability and thus water quality might have to be expected. For example, research suggests that the discharge of the Rhine could increase in winter by the 2080s, by up to 37% from present, but decrease in the summer by 31-35% (Shabalova et al., 2003). Similarly, Krysanova et al. (2005) showed that low flows could decline by 72% in the Elbe by the 2050s.

Hänsel and Matschullat (2006) conducted a trend analysis with observed data for Saxony, in south-east Germany, to determine recent changes in drought patterns. Using the meteorological dry period concept, the authors showed significant trends to longer dry periods during the summer half year.

AVOID programme results

To further quantify the impact of climate change on water stress and the inherent uncertainties, the AVOID programme calculated water stress indices for all countries reviewed in this literature assessment based upon the patterns of climate change from 21

GCMs, following the method described by Gosling et al. (2010) and Arnell (2004). This ensures a consistent methodological approach across all countries and takes consideration of climate modelling uncertainties.

Methodology

The indicator of the effect of climate change on exposure to water resources stress has two components. The first is the number of people within a region with an *increase in exposure to stress*, calculated as the sum of 1) people living in water-stressed watersheds with a significant reduction in runoff due to climate change and 2) people living in watersheds which become water-stressed due to a reduction in runoff. The second is the number of people within a region with a *decrease in exposure to stress*, calculated as the sum of 1) people living in water-stressed watersheds with a significant increase in runoff due to climate change and 2) people living in water-stressed watersheds with a significant increase in runoff due to climate change and 2) people living in watersheds with a significant increase in runoff due to an increase in runoff. It is not appropriate to calculate the net effect of "increase in exposure" and "decrease in exposure", because the consequences of the two are not equivalent. A water-stressed watershed has an average annual runoff less than 1000m³/capita/year, a widely used indicator of water scarcity. This indicator may underestimate water stress in watersheds where per capita withdrawals are high, such as in watersheds with large withdrawals for irrigation.

Average annual runoff (30-year mean) is simulated at a spatial resolution of 0.5x0.5° using a global hydrological model, MacPDM (Gosling and Arnell, 2011), and summed to the watershed scale. Climate change has a "significant" effect on average annual runoff when the change from the baseline is greater than the estimated standard deviation of 30-year mean annual runoff: this varies between 5 and 10%, with higher values in drier areas.

The pattern of climate change from 21 GCMs was applied to MacPDM, under two emissions scenarios; 1) SRES A1B and 2) an aggressive mitigation scenario where emissions follow A1B up to 2016 but then decline at a rate of 5% per year thereafter to a low emissions floor (denoted A1B-2016-5-L). Both scenarios assume that population changes through the 21st century following the SRES A1 scenario as implemented in IMAGE 2.3 (van Vuuren et al., 2007). The application of 21 GCMs is an attempt to quantify the uncertainty due to climate modelling, although it is acknowledged that only one impacts model is applied (MacPDM). Simulations were performed for the years 2030, 2050, 2080 and 2100. Following Warren et al. (2010), changes in the population affected by increasing or decreasing water stress represent the additional percentage of population affected due to climate change, not the absolute change in the percentage of the affected population relative to present day.

Results

The results for Germany are presented in Figure 12. They show that by 2100 under A1B, the median population exposed to increased water stress could be around 40% of the population but uncertainty is high – mitigation reduces this slightly. Conversely, there is higher certainty that few of Germany's population may see a decrease in water stress with climate change. The population exposed to water stress decline by 2100 relative to 2080, due to declining population towards the end of the 21st century.

The projections of increased water stress with climate change contrast with the projections presented by Doll (2009) and Rockström et al. (2009). This might be because the method of calculating water stress for Figure 12 does not explicitly consider groundwater withdrawals. Given that around 70% Germany's available water supply is extracted from groundwater (Aljazzar, 2010, Zhu and Balke, 2008), it is possible that water stress is overestimated in Figure 12, since it does not consider water available from groundwater.

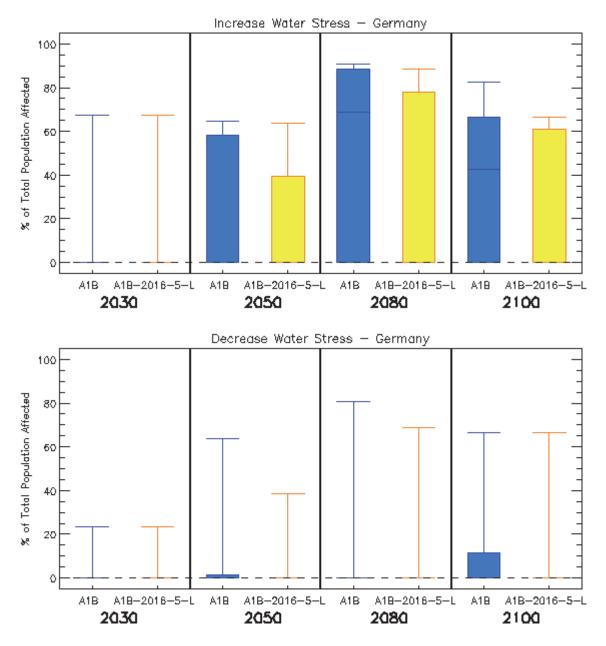


Figure 12. Box and whisker plots for the impact of climate change on increased water stress (top panel) and decreased water stress (bottom panel) in Germany, from 21 GCMs under two emissions scenarios (A1B and A1B-2016-5-L), for four time horizons. The plots show the 25th, 50th, and 75th percentiles (represented by the boxes), and the maximum and minimum values (shown by the extent of the whiskers).

Pluvial flooding and rainfall

Headline

Large-scale modelling studies confirm conclusions from the IPCC AR4 that heavy winter precipitation could increase with climate change for Germany, which might increase the risk of pluvial flooding, depending on changes in rainfall duration and intensity, and the local land cover conditions. However, recent national-scale assessments suggest that increasing trends in precipitation with climate change for Germany are relatively small when compared with emissions uncertainty and that moreover trends of increasing rainfall with climate change may not be significant.

Supporting literature

Introduction

Pluvial flooding can be defined as flooding derived directly from heavy rainfall, which results in overland flow if it is either not able to soak into the ground or exceeds the capacity of artificial drainage systems. This is in contrast to fluvial flooding, which involves flow in rivers either exceeding the capacity of the river channel or breaking through the river banks, and so inundating the floodplain. Pluvial flooding can occur far from river channels, and is usually caused by high intensity, short-duration rainfall events, although it can be caused by lower intensity, longer-duration events, or sometimes by snowmelt. Changes in mean annual or seasonal rainfall are unlikely to be good indicators of change in pluvial flooding; changes in extreme rainfall are of much greater significance. However, even increases in daily rainfall extremes will not necessarily result in increases in pluvial flooding, as this is likely to be dependent on the sub-daily distribution of the rainfall as well as local factors such as soil type, antecedent soil moisture, land cover (especially urbanisation), capacity and maintenance of artificial drainage systems etc. It should be noted that both pluvial and fluvial flooding can potentially result from the same rainfall event.

Assessments that include a global or regional perspective

The IPCC AR4 (2007a) noted that annual precipitation is very likely to increase in most of northern Europe, and daily precipitation extremes are likely to increase under A1B during the 21st century. However, although the simulated responses between models are qualitatively consistent, significant uncertainties remain particularly on the magnitude and geographical details of precipitation change (IPCC, 2007a). The IPCC also noted that the substantial natural variability of European climate is also a major uncertainty, particularly with respect to near-term climate projections. Annual precipitation changes by the end of the 21st century under the A1B emissions scenario range between 0% and 16% in Northern Europe (IPCC, 2007a). The largest increases are simulated during winter. Precipitation extremes during winter are also very likely to increase in both magnitude and frequency (IPCC, 2007a). Note that the majority of Germany lies within the Northern European region as defined in the IPCC AR4 Chapter 11.

Sillmann and Roeckner (2008) find that in northern Europe there is a marked increase of both maximum 5-day rainfall (RX5day) and 95th percentile rainfall (R95p) with climate change under the A1B emissions scenario. RX5day increases by about 9mm until the end of the century, and R95p rises by approximately 90mm. This corresponds to an increase of 17% in RX5day and 64% in R95p. In general, the RX5day differences between A1B and B1 were not pronounced in the northern European region. For R95p, differences in the ensemble means were visible in northern Europe from year 2040 onward but there is a large overlap of the respective ensemble spreads during the whole simulation period. It should be noted, however, that not all of Germany is included in the northern European region, so the results presented by such studies should not be considered as definitive for Germany as a whole.

Similarly, Beniston et al. (2007) found that heavy winter precipitation could increase in central and northern Europe with climate change. These changes were weaker for the B2 emissions scenarios than for the A2 scenario. However, the authors found that model choices can have greater effects on the magnitude (RCM) and pattern (GCM) of response than the choice of scenario. Analysing projections under the A2 and B2 scenarios from an ensemble of RCMs, they found that changes in maximum 5-day rainfall simulated under the B2 scenario were smaller than those simulated under the A2 scenario in two cases, and similar in the other two cases. However, there were no systematic differences in projected increases in maximum 1-day rainfall between the scenarios, though the increases were

positive and up to about 40%. Bates et al. (2008) note that there is a projected increased risk of pluvial, flash floods over the whole of Europe.

National-scale or sub-national scale assessments

Tomassini and Jacob (2009) explored climate change impacts on precipitation for Germany. They observed positive trends in extreme precipitation that match available observations. These trends were, however, relatively small when compared to the uncertainties. The trends became significantly positive to a larger spatial extent only with the SRES A2 emissions scenario simulation (A1B and B1 were also considered). The estimated shape of the extreme value distributions did not change significantly in the scenario simulations compared to the climate model control runs. The most notable feature of the analysis was a reduction of the spatial variance of the trends over Germany in the scenario simulations compared to observations and, in particular, the climate model control runs. Other nationalscale assessments have also demonstrated that trends in heavy precipitation under climate change scenarios for Germany may not be significant (Böhm, 2008, Jacob et al., 2008), which generally supports the results presented by Tomassini and Jacob (2009).

Fluvial flooding

Headline

Climate change could be associated with increases in fluvial flooding for some parts of Germany, particularly the Rhine basin. However, projections of future changes in flood hazard are subject to large uncertainties due to large natural variability as well as large uncertainties in the simulated climate signal. Model simulations have shown diverging responses in extreme river discharge across Germany, mostly associated with the choice of forcing climate model. These conclusions are based upon European-scale modelling studies and national-scale assessments such as the RheinBlick2050 project. Simulations from the AVOID programme, based on 21 GCMs, show a greater tendency towards decreasing flood risk in Germany at first, but later in the century the models become more evenly balanced towards increasing flood risk in the A1B scenario, suggesting large uncertainty. Future assessments should seek to understand, quantify and if possible reduce these uncertainties, for example by a reduction of the biases and by creating probabilistic scenarios based on large ensembles of RCM simulations.

Supporting literature

Introduction

This section summarises findings from a number of post IPCC AR4 assessments on river flooding in Germany to inform and contextualise the analysis performed by the AVOID programme for this project. The results from the AVOID work are discussed in the next section.

Fluvial flooding involves flow in rivers either exceeding the capacity of the river channel or breaking through the river banks, and so inundating the floodplain. A complex set of processes is involved in the translation of precipitation into runoff and subsequently river flow (routing of runoff along river channels). Some of the factors involved are; the partitioning of precipitation into rainfall and snowfall, soil type, antecedent soil moisture, infiltration, land cover, evaporation and plant transpiration, topography, groundwater storage. Determining whether a given river flow exceeds the channel capacity, and where any excess flow will go,

is also not straightforward, and is complicated by the presence of artificial river embankments and other man-made structures for example. Hydrological models attempt to simplify and conceptualise these factors and processes, to allow the simulation of runoff and/or river flow under different conditions. However, the results from global-scale hydrological modelling need to be interpreted with caution, especially for smaller regions, due to the necessarily coarse resolution of such modelling and the assumptions and simplifications this entails (e.g. a 0.5° grid corresponds to landscape features spatially averaged to around 50-55km for mid- to low-latitudes). Such results provide a consistent, high-level picture, but will not show any finer resolution detail or variability. Smaller-scale or catchment-scale hydrological modelling can allow for more local factors affecting the hydrology, but will also involve further sources of uncertainty, such as in the downscaling of global climate model data to the necessary scale for the hydrological models. Furthermore, the application of different hydrological models and analysis techniques often makes it difficult to compare results for different catchments.

Germany has experienced several severe floods in recent years such as catastrophic floods on the Elbe and Danube in August 2002; along tributaries of the Rhine in 2003; and along the Elbe in March and April 2006. In some rivers significant trends in flood indicators have been identified in the streamflow record over the past five decades, particularly in west, south and central Germany (Petrow and Merz, 2009). Consequently many studies have attempted to assess future changes in flood hazard with climate change for Germany, which are discussed below.

Assessments that include a global or regional perspective

Bates et al. (2008) note that in the 2020s there is a projected increased risk of winter floods in northern Europe, The risk of snowmelt flood was projected to shift from spring to winter.

A global modelling study by Hirabayashi et al. (2008), which applied a single GCM under the A1B emissions scenario, projected little change and localised decreases in flood frequency in the next few decades (2001-2030) for Germany. By the end of the century (2071-2100) the simulations showed a mixed pattern of change across the country. However, generally, there was an increase in flood hazard with the flood event that might be expected once every 100 years in the 20th century occurring, on average, once every 30 years by the end of the 21st century. For the Rhine, however, the results indicate an increase in flood frequency of a 100-year flood to more than 300 years, suggesting a significant decrease in flood frequency. It should be noted, however, that results from studies that have applied only a single climate model or climate change scenario should be interpreted with caution. This is

because they do not consider other possible climate change scenarios which could result in a different impact outcome, in terms of magnitude and in some cases sign of change. It is perhaps, for this reason, that the results presented by Hirabayashi et al. (2008) contrast with higher resolution sub-national-scale studies (Görgen et al., 2010, Leipprand et al., 2008, Shabalova et al., 2003).

The Rhine was also included in a global modelling study by Nohara et al. (2006) that applied 19 GCMs under A1B emissions for the end of the 21st century (2081-2100). The results showed diverging changes in the high-flow season discharge of the Rhine depending upon GCM, suggesting large uncertainties predominate in climate change scenarios.

A European-scale study presented by Dankers and Feyen (2008), which applied a very high resolution (~12 km) RCM to drive a flood forecasting model, showed a mixed pattern of changes in the 100-year flood level by the end of the century (2071-2100) under the A2 emissions scenario. In Eastern Germany, particularly in the Oder River basin along the border with Poland, the results showed locally strong increases up to 40% higher than the current 100-year discharge level. In central Germany there were, however, localised decreases while in the Rhine there was mostly little change. For the River Elbe at Dresden the simulations showed a strong increase in extreme discharges particularly at higher return periods (more than 10 years). Based on the same high resolution climate scenarios but using a more detailed flood model setup, Dankers et al. (2007) found comparable results in the Upper Danube basin with increases in extreme discharge levels particularly with higher return periods (more than 30 years). However, in some rivers originating from the Alps such as the Lech and Inn rivers the simulations suggested a decrease in extreme river flows.

In a follow-up study that applied two RCMs with two emission scenarios (SRES A2 and B2), Dankers and Feyen (2009) found diverging responses in extreme river discharge across Germany, which was found to be associated mostly with the choice of climate model. Dankers and Feyen (2009) also found that some of the changes in simulated flood hazard can partly be attributed to large, decadal-scale variability in the simulated climate and can be expected to occur naturally when comparing two 30-year time periods, even without a change in greenhouse gas forcing. At the scale of individual river basins, using a different combination of climate models or assuming a different emissions scenario sometimes resulted in a very different or even opposite climate change signal in flood hazard. Nevertheless, some changes were found to be consistent among at least a majority of the model simulations, such as a considerable increase in extreme discharge in the Oder River and a consistent decrease in flood hazard in parts of the Weser catchment. As highlighted by Dankers and Feyen (2009), projections of future changes in flood hazard are subject to large uncertainties due to large natural variability and large uncertainties in the simulated climate signal. Future assessments should focus on a reduction of the biases and on creating probabilistic scenarios based on larger ensembles of RCM simulations, but should also address hydrological modelling uncertainty.

National-scale or sub-national scale assessments

There have been several climate change impact studies conducted for individual river catchments in Germany. For example in the Elbe basin, Hatterman et al. (2008) projected a shift in the occurrence of flood events from early spring to early winter with climate change due to less retention of runoff in snow. In a modelling study of the Rhine basin, Hurkmans et al. (2010) found that simulated annual maximum river flows at nearly all return periods were generally higher than in the reference period (1950-2000) under three different emission scenarios (A2, A1B and B1). In the most extreme scenarios, an event with the magnitude of the most extreme flooding events in the last half century occurred on average every 5-6 years in the future. Likewise, Te Linde et al. (2010) projected a basin-wide increase in peak discharge by 2050 of 8%-17% for discharge levels with probabilities between 1 in 10 and 1 in 1250 years. In a related study, Te Linde et al. (2011) projected that by 2030 the annual expected damage from flooding over the entire Rhine basin may increase by between 54% and 230%, of which the major part (around 75%) can be accounted for by climate change.

The RheinBlick2050 project (Görgen et al., 2010) assessed the impacts of future climate change on discharge of the Rhine River and its major tributaries. The project applied an ensemble of 20 dynamically downscaled transient bias-corrected regional climate simulations (control runs and projections) as forcing data for the HBV134 hydrological model at a daily temporal resolution over the catchments of the Rhine River. The authors of the report note that discharge projections presented here give an indication of the changes in the hydrometeorological system of the Rhine River and its catchments as a result of climate change projections that are regarded the state-of-the-art in 2010. For high flow statistics, Görgen et al. (2010) concluded that overall tendencies to increase are projected for Raunheim (Main River), Trier (Moselle River), Köln and Lobith, by the 2080s. For the 2030s, the tendencies are generally smaller (except for Raunheim) or absent. The authors noted that the relative uncertainties become larger going from the 2030s to the 2080s. In addition the uncertainties increased going from moderately extreme flows to highly extreme flows. Görgen et al. (2010) drew no conclusions for gauges with a flow regime characterised by summer high flows, like Basel and Maxau, since there is limited confidence in the extreme

discharge projections. The results presented by Görgen et al. (2010) generally support conclusions from Leipprand et al. (2008) and the larger-scale study presented by Dankers and Feyen (2009); both studies found diverging responses in extreme river discharge across Germany, which were associated mostly with the choice of climate model.

Similar conclusions on seasonal changes in discharge with climate change are also noted in a review by Leipprand et al. (2008). The authors found that while Germany might face more floods in the Rhine basin with climate change, particularly in winter, reductions in average river discharges might be experienced for the summer months. Shabalova et al. (2003) found that the discharge of the Rhine could increase in winter by the 2080s, by up to 37% from present.

AVOID programme results

To quantify the impact of climate change on fluvial flooding and the inherent uncertainties, the AVOID programme calculated an indicator of flood risk for all countries reviewed in this literature assessment based upon the patterns of climate change from 21 GCMs (Warren et al., 2010). This ensures a consistent methodological approach across all countries and takes consideration of climate modelling uncertainties.

Methodology

The effect of climate change on fluvial flooding is shown here using an indicator representing the percentage change in average annual flood risk within a country, calculated by assuming a standardised relationship between flood magnitude and loss. The indicator is based on the estimated present-day (1961-1990) and future flood frequency curve, derived from the time series of runoff simulated at a spatial resolution of $0.5^{\circ}x0.5^{\circ}$ using a global hydrological model, MacPDM (Gosling and Arnell, 2011). The flood frequency curve was combined with a generic flood magnitude–damage curve to estimate the average annual flood damage in each grid cell. This was then multiplied by grid cell population and summed across a region, producing in effect a population-weighted average annual damage. Flood damage is thus assumed to be proportional to population in each grid cell, not the value of exposed assets, and the proportion of people exposed to flood is assumed to be constant across each grid cell (Warren et al., 2010).

The national values are calculated across major floodplains, based on the UN PREVIEW Global Risk Data Platform (preview.grid.unep.ch). This database contains gridded estimates, at a spatial resolution of 30 arc-seconds (0.00833°x0.00833°), of the estimated frequency of flooding. From this database the proportion of each 0.5°x0.5° grid cell defined as floodplain was determined, along with the numbers of people living in each 0.5°x0.5° grid cell in floodprone areas. The floodplain data set does not include "small" floodplains, so underestimates actual exposure to flooding. The pattern of climate change from 21 GCMs was applied to MacPDM, under two emissions scenarios; 1) SRES A1B and 2) an aggressive mitigation scenario where emissions follow A1B up to 2016 but then decline at a rate of 5% per year thereafter to a low emissions floor (denoted A1B-2016-5-L). Both scenarios assume that population changes through the 21st century following the SRES A1 scenario as implemented in IMAGE 2.3 (van Vuuren et al., 2007). The application of 21 GCMs is an attempt to quantify the uncertainty due to climate modelling, although it is acknowledged that only one impacts model is applied (MacPDM). Simulations were performed for thirty-year periods centred on the years 2030, 2050, 2080 and 2100. The result represents the change in flood risk due to climate change, not the change in flood risk relative to present day (Warren et al., 2010).

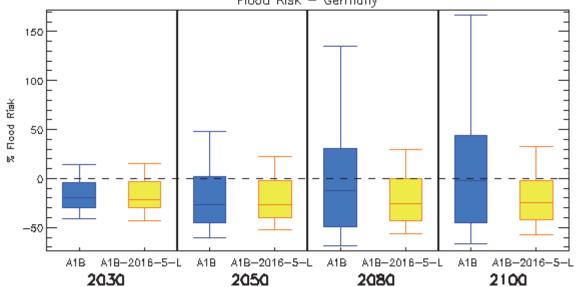
Results

The results for Germany are presented in Figure 23. By the 2030s, the models project a range of changes in mean fluvial flooding risk over Germany in both scenarios, with some models projecting decreases and others increases. However, the balance is more towards lower flood risk, with more than 75% of models projecting a decrease. The largest decrease projected for the 2030s is a 40% decline in the average annual flood risk, and the largest increase is 15%. The mean across all projections for this time period is a decrease in flood risk of approximately 20%.

By 2100 the model projections become more evenly balanced between decreased and increased flood risk, and the difference in projections from the different models becomes greater. Both these aspects of the results are more pronounced for the A1B scenario than the mitigation scenario. Under the mitigation scenario, three quarters of the models still project a lower flood risk (down to approximately -60%). The mean of all projections by the end of the century for the aggressive mitigation scenario is a 25% decrease, and the upper projection is an increase of approximately 30%. Under the A1B scenario, about half the models project a decreased flood risk (down to -60%), and half project and increase (up to

approximately 170%). However, the mean projection for the A1B scenario at the end of the century is for no change in the annual average flood risk.

So for Germany, the models show a greater tendency towards decreasing flood risk at first in both scenarios, but later in the century half of the models indicate an increasing flood risk in the A1B scenario. The differences between the model projections are greater later in the century and particularly for A1B. The high uncertainty in the projections, supports conclusions from several other studies, which note large uncertainty associated with the estimation of future flood risk in Germany with climate change (Görgen et al., 2010, Leipprand et al., 2008, Dankers and Feyen, 2009). However, even under the A1B scenario about half of the models in Figure 13 project a decrease in flood risk in Germany, which is somewhat in contrast to previous studies (Görgen et al., 2010, Leipprand et al., 2008, Dankers and Feyen, 2009). This could be due to a number of reasons. For example, the calculations applied here and shown in Figure 13 include population data, whereas most other studies are focussed purely on hydrological change. However, differences could also arise due to the choice of climate scenarios, how the scenarios were applied or even the hydrological models that were used.



Flood Risk — Germany

Figure 13. Box and whisker plots for the percentage change in average annual flood risk within Germany, from 21 GCMs under two emissions scenarios (A1B and A1B-2016-5-L), for four time horizons. The plots show the 25th, 50th, and 75th percentiles (represented by the boxes), and the maximum and minimum values (shown by the extent of the whiskers).

Tropical cyclones

This country is not impacted by tropical cyclones.

Coastal regions

Headline

Germany has coasts on both the North and Baltic Seas extending over five coastal states. Large-scale and national-scale studies suggest that without adaptation, the number of people at risk from coastal flooding in Germany due to Sea Level Rise (SLR) under climate change scenarios could be around 300,000. However, these impacts could be reduced substantially by adaptation and climate change mitigation measures. One study demonstrates that while Germany is potentially highly threatened by SLR, adaptation in the form of raising of flood dykes and the application of beach nourishment could greatly reduce these impacts to levels which appear manageable. These studies add further detail to knowledge documented in the IPCC AR4.

Supporting literature

Assessments that include a global or regional perspective

The IPCC AR4 concluded that at the time, understanding was too limited to provide a best estimate or an upper bound for global SLR in the twenty-first century (IPCC, 2007b). However, a range of SLR, excluding accelerated ice loss effects was published, ranging from 0.19m to 0.59m by the 2090s (relative to 1980-2000), for a range of scenarios (SRES A1FI to B1). The IPCC AR4 also provided an illustrative estimate of an additional SLR term of up to 17cm from acceleration of ice sheet outlet glaciers and ice streams, but did not suggest this is the upper value that could occur. Although there are published projections of SLR in excess of IPCC AR4 values (Nicholls et al., 2011), many of these typically use semi-empirical methods that suffer from limited physical validity and further research is required to produce a more robust estimate. Linking sea level rise projections to temperature must also be done with caution because of the different response times of these two climate variables to a given radiative forcing change.

Nicholls and Lowe (2004) previously showed that mitigation alone would not avoid all of the impacts due to rising sea levels, adaptation would likely be needed too. Recent work by van Vuuren et al. (2011) estimated that, for a world where global mean near surface temperatures reach around 2°C by 2100, global mean SLR could be 0.49m above present levels by the end of the century. Their sea level rise estimate for a world with global mean temperatures reaching 4°C by 2100 was 0.71m, suggesting around 40% of the future increase in sea level to the end of the 21st century could be avoided by mitigation. A qualitatively similar conclusion was reached in a study by Pardaens et al. (2011), which examined climate change projections from two GCMs. They found that around a third of global-mean SLR over the 21st century could potentially be avoided by a mitigation scenario under which global-mean surface air temperature is near-stabilised at around 2°C relative to pre-industrial times. Under their baseline business-as-usual scenario the projected increase in temperature over the 21st century is around 4°C, and the sea level rise range is 0.29-0.51m (by 2090-2099 relative to 1980-1999; 5% to 95% uncertainties arising from treatment of land-based ice melt and following the methodology used by the IPCC AR4). Under the mitigation scenario, global mean SLR in this study is projected to be 0.17-0.34m.

The IPCC 4th assessment (IPCCa) followed Nicholls and Lowe (2004) for estimates of the numbers of people affected by coastal flooding due to sea level rise. Nicholls and Lowe (2004) projected for the North and West Europe region that an additional 100 thousand people per year could be flooded due to sea level rise by the 2080s relative to the 1990s for the SRES A2 Scenario (note this region also includes other countries, such as UK and Norway). For the Baltic region, this figure is less than 100,000 people per year. However, it is important to note that this calculation assumed that protection standards increased as GDP increased, although there is no additional adaptation for sea level rise. More recently, Nicholls et al. (2011) also examined the potential impacts of sea level rise in a scenario that gave around 4°C of warming by 2100. Readings from Figure 3 from Nicholls et al. (2011) for the North and West Europe region suggest that less than an approximate one million additional people could be flooded for a 0.5 m SLR (assuming no additional protection), with less than 1 million addition people flooded in the Baltic region. Nicholls et al. (2011) also looked at the consequence of a 2m SLR by 2100, however as we consider this rate of SLR to have a low probability we don't report these figures here.

The European Commission (2009) assessed the vulnerability of several European countries to SLR. The study showed that of Germany's coastline with the North Sea, 70% of this comprises 10km long stretches that are below 5m elevation and that 226km is subject to

erosion. The study also calculated that 5% of GDP is located within 50km of the coast and that 7% of the country's population live within this zone.

Tol et al. (2008) note that there is a high awareness of the potential impact of SLR in Germany, particularly that associated with the Baltic Sea. Tol et al. (2008) suggests that SLR from the Baltic Sea could have potentially large impacts on the coastal zone, but with limited consequences for people and economy.

Recent results from the PESETA (Projection of Economic impacts of climate change in Sectors of the European Union based on boTtom-up Analysis) project have afforded consistent quantitative projections of the impact of SLR for several European countries (Richards and Nicholls, 2009). These are advantageous because previous European assessments have tended to be more qualitative in nature (Nicholls, 2000). The results show that while Europe is potentially highly threatened by SLR, adaptation (in the form of the two protection options considered) can greatly reduce these impacts to levels which appear manageable. The adaptation methods and costs assessed were the raising of flood dykes and the application of beach nourishment. Richards and Nicholls (2009) show that there are almost immediate benefits of adaptation, and the analysis suggests that widespread adaptation to SLR across Europe would be prudent. The assessment considered SLR projections from two GCMs, ECHAM4 and HadCM3. For each of these, SLR estimates for low, medium and high climate sensitivities were applied, and under the A2 and B2 emissions scenarios. To further quantify uncertainty, the upper and lower estimates of global SLR from the IPCC TAR were also applied. The estimates of global SLR considered by Richards and Nicholls (2009) are summarised in Table 10. Given that the IPCC TAR (IPCC, 2001) estimates of SLR encompass the full range of uncertainty that Richards and Nicholls (2009) considered, impacts for the IPCC TAR low and high scenarios are presented in Table 11. The results show that by the 2080s under the high SLR scenario and without adaptation, the average annual number of people flooded is around 293,000. This is greatly reduced with adaptation, to around 2,900. Under the low SLR scenario, 4,400 people are flooded annually without adaptation and 2,100 are flooded with adaptation. The results highlight the importance of climate sensitivity in determining the impacts as well as demonstrating clear potential benefits of adaptive measures, which by the 2080s can almost completely remove any incremental climate change effect.

GCM	EC	HAM4	Had	ICM3	IPCC TAR
SRES scenario	A2	B2	A2	B2	A2/B2
Climate sensitivity					
Low	29	22	25	19	9
Medium	43	36	40	34	
High	58	50	56	48	88

Table 10. Global SLR (cm) for low, medium and high climate sensitivities at 2100, for the A2 and B2 SRES scenarios, that were applied by Richards and Nicholls (2009)

		Docolino	IPCC A2 202	IPCC A2 2020s Low SLR	IPCC A2 202	IPCC A2 2020s High SLR	IPCC A2 208	IPCC A2 2080s Low SLR	IPCC A2 208	IPCC A2 2080s High SLR
Country	Indicator	(1995)	No adaptation	With adaptation	No adaptation	With adaptation	No adaptation	With adaptation	No adaptation	With adaptation
	Total damage costs (millions €/year)	253.2	362.7	228.8	434.9	291.7	673.3	220.9	3359.8	456.6
	Land loss (submergence) (km ² /year)	0	0.2	0.2	0.6	0.6	0.2	0.2	130.8	1.6
	Migration due to land loss (1000s of people/year)	0	0	0	0	0	0	0	19.5	0
	Net loss of wetland area (km²/year)	12.4	300.7	300.7	688.7	688.7	950.9	950.9	1963.8	1963.8
	People actually flooded (1000s/year)	1.96	2.4	2.0	2.9	2.1	3.0	1.8	463.7	2.5
	Salinity intrusion costs (millions €/year)	135.85	145.3	145.3	153.5	153.5	199.9	199.9	268.7	268.7
	Total damage costs (millions €/year)	428.1	473.7	398.9	590.3	448.3	782.2	344.6	2607.6	686.9
	Land loss (submergence) (km ² /year)	0	0	0	0	0	0	0	178.1	0
	Migration due to land loss (1000s of people/year)	0	0	0	0	0	0	0	14.3	0.1
Germany	Net loss of wetland area (km²/year)	0.3	146.4	146.4	418.3	418.3	673.2	673.2	1938.5	1938.5
	People actually flooded (1000s/year)	2.23	2.6	2.3	3.7	2.5	4.4	2.1	293.3	2.9
	Salinity intrusion costs (millions €/year)	274.78	240.8	240.8	251.2	251.2	343.3	343.3	431.4	431.4
	Total damage costs (millions €/year)	8.5	18.5	6.6	271.9	7.1	444.2	9.6	8044.2	17.0
	Land loss (submergence) (km²/year)	0	0	0	0	0	0	0	188.5	0.1
	Migration due to land loss (1000s of people/year)	0	0	0	0	0	0	0	55.2	0
Italy	Net loss of wetland area (km²/year)	9.6	78.2	78.2	155.5	155.5	221.8	221.8	393.7	393.7
	People actually flooded (1000s/year)	1.49	1.9	1.5	3.1	1.6	3.3	1.5	512.9	2.3
	Salinity intrusion costs (millions €/year)	5.84	6.4	6.4	6.7	6.7	9.2	9.2	11.9	11.9

	Total damage costs (millions €/year)	15.3	23.6	9.4	55.6	14.5	75.6	3.9	749.7	32.0
	Land loss (submergence) (km²/year)	0	0	0	0	0	0	0	38.3	0.1
<u></u>	Migration due to land loss (1000s of people/year)	0	0	0	0	0	0	0	7.0	0
Spall	Net loss of wetland area (km²/year)	0.2	64.4	64.4	174.3	174.3	242.5	242.5	659.8	659.8
	People actually flooded (1000s/year)	0.73	6.0	0.7	1.3	0.8	1.1	2.0	321.8	1.1
	Salinity intrusion costs (millions €/year)	0.92	1.0	1.0	1.0	1.0	1.4	1.4	1.8	1.8
	Total damage costs (millions €/year)	189.5	414.2	50.1	743.5	136.4	1087.0	6.3	10062.6	155.1
	Land loss (submergence) (km²/year)	0	0	0	0	0	0	0	239.3	0
Ì	Migration due to land loss (1000s of people/year)	0	0	0	0	0	0	0	73.6	0
5	Net loss of wetland area (km²/year)	8.5	175.1	175.1	522.6	522.6	789.4	789.4	1690.2	1690.2
	People actually flooded (1000s/year)	4.85	5.7	4.4	6.1	4.9	6.8	4.3	986.3	5.6
	Salinity intrusion costs (millions €/year)	2.32	2.5	2.5	2.8	2.8	3.6	3.6	5.7	5.7
Table 11	Table 11. Estimates of the impact of SLR with the low and	SLR with th		high estimates of SLR from the IPCC TAR in the 2020s and 2080s, and assuming no adaptation or	SLR from the	IPCC TAR in	the 2020s and	d 2080s, and ¿	assuming no a	adaptation or

25 ŗ adaptation. Baseline impacts (1995) are also presented. Data is from Richards and Nicholls (2009). Hanson et al. (2010) present a global-scale analysis of the impact of SLR on coasts and include estimates for port cities in Germany. The study investigated population exposure to global SLR, natural and human subsidence/uplift, and more intense storms and higher storm surges, for 136 port cities across the globe. Future city populations were calculated using global population and economic projections, based on the SRES A1 scenario up to 2030. The study accounted for uncertainty on future urbanization rates, but estimates of population exposure were only presented for a rapid urbanisation scenario, which involved the direct extrapolation of population from 2030 to 2080. All scenarios assumed that new inhabitants of cities in the future could have the same relative exposure to flood risk as current inhabitants. The study is similar to a later study presented by Hanson et al. (2011) except here, different climate change scenarios were considered, and published estimates of exposure are available for more countries, including Germany. Future water levels were generated from temperature and thermal expansion data related to greenhouse gas emissions with SRES A1B (un-mitigated climate change) and under a mitigation scenario where emissions peak in 2016 and decrease subsequently at 5% per year to a low emissions floor (2016-5-L). Table 12 shows the aspects of SLR that were considered for various scenarios and Table 13 displays regional population exposure for each scenario in the 2030s, 2050s and 2070s. The results show that Germany is second only to the UK in Europe for countries with port cities most impacted by SLR, and around the midpoint of all countries considered globally (Table 13). The effect of climate change is observed by comparing the projections in Table 13 with the estimates for exposure in the absence of climate change that are presented in Table 14. At present, 261,000 people in port cities are exposed to SLR in Germany. By the 2070s in the absence of climate change 280,000 are exposed. With climate change in the 2070s, 309,000 people in port cities are exposed. Hanson et al. (2010) also demonstrated that aggressive mitigation scenario could avoid an exposure of around 15,000 people in German port cities, relative to un-mitigated climate change (see Table 14). Estimates presented by Nicholls and Mimura (1998) are similar to the projections estimated by Hanson et al. (2010) for Germany. Nicholls and Mimura (1998) calculated that a 1m SLR for Germany would be associated with 309,000 people exposed to flooding, a land loss of 13,900km² (this amount of land loss is also estimated by Sterr and Simmering (1996)), and a capital value loss of \$7,500m (0.05% of GDP).

S	Scenario			Wate	r levels	
			Climat	te		Subsidence
Code	Description	More intense storms	Sea- level change	Higher storm surges	Natural	Anthropogenic
FNC	Future city	V	х	Х	х	Х
FRSLC	Future City Sea-Level Change	v	v	x	v	х
FCC	Future City Climate Change	v	v	v	v	x
FAC	Future City All Changes	V	v	v	V	v

Table 12. Summary of the aspects of SLR considered by Hanson et al. (2010). 'V' denotes that the aspect was considered in the scenario and 'x' that it was not.

		sction	FNC	18,600	13,900	10,700	5,070	1,530	864	569	489	303	280	169	175	29	27	18	4	0		
	2070	Water level projection	FRSL C	20,800	15,600	12,100	5,580	1,830	940	640	545	325	290	197	186	38	30	23	9	0		
		Water	FCC	22,600	17,900	12,700	5,970	1,830	940	716	585	377	309	226	196	38	30	23	9	0		
			FAC	27,700	20,600	12,800	7,800	2,680	940	716	614	377	309	226	196	38	30	23	9	0		
			Ports	15	9	17	9	4	10	2	2	3	٢	٢	5	٢	2	٢	٢	0		
			Country	CHINA	VIDNI	SN	NAPAN	VISENDONI	BRAZIL	NU	CANADA	REPUBLIC OF KOREA	GERMANY	RUSSIA	AUSTRALIA	SAUDI ARABIA	SOUTH AFRICA	FRANCE	ΙΤΑLΥ	MEXICO		
		Water level projection	jection	FNC	17,400	12,500	9,970	4,760	1,500	879	521	457	318	269	173	181	27	27	17	3	0	
tion			FRSLC	18,700	13,600	10,800	5,000	1,610	929	564	486	341	273	173	191	33	28	19	4	0		
on project				FCC	19,700	14,600	11,200	5,280	1,610	929	609	512	361	287	202	197	33	28	19	4	0	
Rapid urbanisation projection	2050		FAC	23,000	16,400	11,300	6,440	2,110	929	609	549	361	287	202	197	33	28	19	4	0		
apid נ			Ports	15	9	17	9	4	10	2	2	3	1	1	5	1	2	۱	1	0		
Ľ			Country	CHINA	VIDNI	SN	NAPAN	INDONESIA	BRAZIL	ЯN	CANADA	REPUBLIC OF KOREA	GERMANY	RUSSIA	AUSTRALIA	SAUDI ARABIA	SOUTH AFRICA	FRANCE	ΙΤΑLΥ	MEXICO		
		Water level projection			FNC	14,600	9,970	8,460	4,390	1,170	802	459	405	441	248	177	157	29	22	15	2	0
			FRSL C	15,400	10,300	8,830	4,430	1,200	833	478	422	331	253	177	157	30	24	15	2	0		
			level proj	FCC	15,500	10,800	8,960	4,610	1,200	833	497	433	344	257	177	162	30	24	15	2	0	
	2030	Water	FAC	17,100	11,600	8,990	5,260	1,420	833	497	459	344	257	177	162	30	24	15	2	0		
			Ports	15	9	17	9	4	10	2	2	3	Ļ	1	5	2	-	٢	1	0		
			Country	CHINA	VIDN	SN	JAPAN	INDONESIA	BRAZIL	NU	CANADA	REPUBLIC OF KOREA	GERMANY	RUSSIA	AUSTRALIA	SOUTH AFRICA	SAUDI ARABIA	FRANCE	ΙΤΑLΥ	MEXICO		

Table 13. National estimates of port city population exposure (1,000s) for each water level projection (ranked according to exposure with the FAC (Future City All Changes) scenario) under a rapid urbanisation projection for the 2030s, 2050s and 2070s. Estimates for present day exposure and in the absence of climate change (for 2070 only) for comparison are presented in Table 14. Data is from Hanson et al. (2010) and has been rounded down to three significant figures.

			Pc	Population exposure		
			2070. Rapid	2070. Rapid urbanisation, FAC water level scenario	ater level scenario	
Country	Ports	Current	No climate change	A1B un-mitigated	Mitigated (2016-5-L)	Exposure avoided
CHINA	15	8,740	18,600	27,700	26,500	1,140
UNITED STATES	17	6,680	10,700	12,800	12,300	505
RUSSIA	1	189	169	226	197	28
JAPAN	9	3,680	5,070	7,800	7,290	515
SOUTH AFRICA	2	54	27	30	56	0
INDIA	9	5,540	13,900	20,600	18,900	1,670
BRAZIL	10	222	864	940	926	14
MEXICO	0	0	0	0	0	0
CANADA	2	308	489	614	669	15
AUSTRALIA	5	66	175	196	190	9
INDONESIA	4	602	1,530	2,680	2,520	156
REPUBLIC OF KOREA	3	294	303	377	343	34
UK	2	414	569	716	665	51
FRANCE	٢	13	18	23	20	2
ΙΤΑLΥ	L	2	4	9	9	0
GERMANY	1	261	280	309	295	15
SAUDI ARABIA	L	15	29	38	35	£
		(1 000c) in precent (current)		and in the 2070c in the absence of climate change (no climate change)	sence of climate chan	ac /nc climate change

Table 14. Exposed port city population (1,000s) in present (current), and in the 2070s in the absence of climate change (no climate change), with unmitigated climate change (model and the climate change), with unmitigated climate change (A1B un-mitigated), and mitigated climate change (mitigated 2016-5-L), under the rapid urbanisation and FAC (Future City All Changes) water level scenarios. The final column shows the potential avoided exposure, as a result of mitigation. Data is from Hanson et al. (2010) and has been rounded down to three significant figures.

To further quantify the impact of SLR and some of the inherent uncertainties, the DIVA model was used to calculate the number of people flooded per year for global mean sea level increases (Brown et al., 2011). The DIVA model (DINAS-COAST, 2006) is an integrated model of coastal systems that combines scenarios of water level changes with socio-economic information, such as increases in population. The study uses two climate scenarios; 1) the SRES A1B scenario and 2) a mitigation scenario, RCP2.6. In both cases an SRES A1B population scenario was used. The results are shown in Table 15.

	A1B		RCP	
	Low	High	Low	High
Additional people flooded (1000s)	2.06	19.98	1.74	6.96
Loss of wetlands area (% of country's total wetland)	31.18%	37.97%	27.12%	35.83%

Table 15. Number of additional people flooded (1000s), and percentage of total wetlands lost by the 2080s under the high and low SRES A1B and mitigation (RCP 2.6) scenarios (Brown et al., 2011).

National-scale or sub-national scale assessments

Sterr (2008) estimated that under a 1m accelerated SLR scenario, the recurrence of devastating storm floods in Germany that presently have a probability of 1 in 100 could decrease to a 1 in 10 or even 1 in 1 probability. Sterr (2008) performed vulnerability assessments in Germany at three scales: (i) the national level, i.e., for all coastal areas lying below 5 m (Baltic Sea Coast) and 10 m (North Sea Coast), (ii) the regional level for the coastal state of Schleswig-Holstein, and (iii) the local level for selected communities within this state. When Sterr (2008) compared the findings from these analyses, the results showed that the economic risks of flooding and erosion were highest when detailed studies covering the full range of infrastructure assets were used. However, the actual risk areas in detailed studies may be more confined when considering local topography and infrastructure such as road dams. Sterr (2008) found that nationally, an accelerated SLR of 1m could put more than 300,000 people at risk in the coastal cities and communities. This scale of impact is comparable to that presented by Richards and Nicholls (2009), Hanson et al. (2010) and Nicholls and Mimura (1998) for Germany.

The Final Report of the Trilateral Working Group on Coastal Protection and Sea Level Rise (CPSL, 2001) assessed the impact of SLR on the coastal region of Germany by the Wadden

Sea. The report demonstrated that flooding time of intertidal flats could increase by 2.5-7.5% under a 25cm SLR scenario, and between 5-15% under a 50cm scenario. CPSL (2001) showed that this increase could be further aggravated if storminess increases as well. However, the report noted that the changes could strongly differ from one tidal basin to another, depending on the amount of sediment available to balance future SLR. Indeed, some tidal basins already suffer from a sediment deficit, meaning that the flooding time in these tidal basins could be much more affected than suggested by the mean values presented. The report also showed that the effects of SLR on the surface area of the intertidal flats were similar as those for the flooding time. The spatial extent might, under the 50cm SLR scenario, diminish by up to 15% or (720 km²), compared to the present situation (4,800 km²).

The European-scale assessment Richards and Nicholls (2009) highlighted the importance of adaptation in Germany, if major SLR impacts are to be avoided. Ongoing research by the Nordwest2050 project demonstrates that adaptation options like second dike lines in the hinterland, polder or retention areas, floating homes, and integrated coastal protection are adaptation measures considered by Germany to adapt to future SLR (Schuchardt et al., 2010, Schuchardt et al., 2011). In Germany, capital coastal protection measures are open to co-financing of up to 70% by the federal government, whereas the maintenance of existing structures is financed 100% by the respective state (European Commission, 2009). The involvement and role of the local water boards differs from state to state. In Hamburg, the water boards do not play any role in coastal protection whereas in Schleswig-Holstein, water boards are responsible for regional dikes at the mainland coast of the Baltic Sea and for the second dike line at the North Sea coast and the state dikes are state responsibility (European Commission, 2009). Also in Niedersachsen the state government and water boards work together.

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Acknowledgements

Funding for this work was provided by the UK Government Department of Energy and Climate Change, along with information on the policy relevance of the results.

The research was led by the UK Met Office in collaboration with experts from the University of Nottingham, Walker Institute at the University of Reading, Centre for Ecology and Hydrology, University of Leeds, Tyndall Centre — University of East Anglia, and Tyndall Centre — University of Southampton.

Some of the results described in this report are from work done in the AVOID programme by the UK Met Office, Walker Institute at the University of Reading, Tyndall Centre — University of East Anglia, and Tyndall Centre — University of Southampton.

The AVOID results are built on a wider body of research conducted by experts in climate and impact models at these institutions, and in supporting techniques such as statistical downscaling and pattern scaling.

The help provided by experts in each country is gratefully acknowledged — for the climate information they suggested and the reviews they provided, which enhanced the content and scientific integrity of the reports.

The work of the independent expert reviewers at the Centre for Ecology and Hydrology, University of Oxford, and Fiona's Red Kite Climate Consultancy is gratefully acknowledged.

Finally, thanks go to the designers, copy editors and project managers who worked on the reports.

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