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How reliable are projections of future flood damage?

P. Bubeck1,2, H. de Moel1, L. M. Bouwer1, and J. C. J. H. Aerts1

1Institute for Environmental Studies, Faculty of Earth and Life Sciences, VU University Amsterdam, The Netherlands
2German Research Centre for Geosciences GFZ, Helmholtz Centre Potsdam, Section Hydrology, Germany

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Abstract. Flood damage modelling is an important component in flood risk management, and several studies have investigated the possible range of flood damage in the coming decades. Generally, flood damage assessments are still characterized by considerable uncertainties in stage-damage functions and methodological differences in estimating exposed asset values. The high variance that is commonly associated with absolute flood damage assessments is the reason for the present study that investigates the reliability of estimates of relative changes in the development of potential flood damage. While studies that estimate (relative) changes in flood damage over time usually address uncertainties resulting from different projections (e.g. land-use characteristics), the influence of different flood damage modelling approaches on estimates of relative changes in the development of flood damage is largely unknown. In this paper, we evaluate the reliability of estimates of relative changes in flood damage along the river Rhine between 1990 and 2030 in terms of different flood-damage modelling approaches. The results show that relative estimates of flood damage developments differ by a factor of 1.4. These variations, which result from the application of different modelling approaches, are considerably smaller than differences between the approaches in terms of absolute damage estimates (by a factor of 3.5 to 3.8), or than differences resulting from land-use projections (by a factor of 3). The differences that exist when estimating relative changes principally depend on the differences in damage functions. In order to improve the reliability of relative estimates of changes in the development of potential flood damage, future research should focus on reducing the uncertainties related to damage functions.

1 Introduction

In recent years, flood management throughout Europe has gradually shifted from what are called “flood control approaches” to more integrated concepts, referred to as “flood risk management”. While flood control approaches predominantly focused on preventing flood events with specific predefined return periods, flood risk management also takes into account the expected consequences of flooding such as direct economic losses or loss of life (Büchele et al., 2006; de Moel et al., 2009; Merz et al., 2010). Risk in this context is defined as “probability times damage”, and thus describes the expected damage that can occur or will be exceeded with a certain probability in a certain period (e.g. Merz et al., 2010).

Following this shift to risk-based approaches, there has been an increasing interest in flood impact assessment and especially the estimation of direct economic losses (Dutta et al., 2003; Merz et al., 2004; Hall et al., 2005; Penning-Rossell et al., 2005; Thielen et al., 2005; Bouwer et al., 2009; Luino et al., 2009; Kreibich et al., 2010). Knowledge of potential flood damage has a great importance for, inter alia, the identification of people and assets at risk, the planning and evaluation of effective flood mitigation and control measures, the creation of flood risk maps for awareness raising, and the calculation of flood insurance premiums (Messner et al., 2007; Merz et al., 2010). Furthermore, flood damage is projected to increase in the coming decades owing to on-going development in flood-prone areas and the projected effects of climate change on river discharges, and consequently flood probabilities (Middelkoop et al., 2001; IPCC, 2007; te Linde et al., 2010). Against this background, a growing number of studies have estimated the range of possible changes in the development of future flood damage in Europe (Hall et al., 2005; Aerts et al., 2008; ABI, 2009; Feyen et al., 2009; Maaskant et al., 2009; Bouwer et al., 2010; te Linde et al., 2011).

Generally, flood damage assessments are still characterized by significant uncertainties associated with
stage-damage functions, as well as methodological differences in estimating the exposed asset values linked to these curves (Merz et al., 2004; Apel et al., 2008, 2009; Freni et al., 2010; Merz et al., 2010; de Moel and Aerts, 2011). While further efforts are currently being undertaken to reduce the uncertainties of flood loss assessment, it can be assumed that considerable uncertainty will remain in the coming years (Jonkman et al., 2008; Maaskant et al., 2009; Kreibich et al., 2010; Elmer et al., 2010). In order to better manage the large variations that are commonly associated with assessments of absolute flood damage, it is suggested that it would be useful to investigate the reliability of estimates of relative changes in the development of potential flood damage, especially in terms of the differences stemming from flood damage modelling approaches (de Moel and Aerts, 2011). Gaining insights into the reliability of relative estimates (as the percentage change of a reference situation) of flood damage in scenario studies is important, as these often form the basis of decision making and are used, for example, to evaluate the effectiveness of various risk reducing-strategies (ICPR, 2006; Aerts et al., 2008; ABI, 2009; Bouwer et al., 2010). The latter purpose is especially important because many investments in flood control and mitigation measures take 20 to 30 yr to design, plan and implement and are also designed for long life spans (see e.g. Hallegatte, 2009; Dircke et al., 2010).

Furthermore, insights into the influence of different flood damage modelling approaches on relative estimates of flood damage developments are needed, given the increasing emphasis on basin-wide approaches for trans-boundary flood risk assessment. A basin-wide approach is currently required in Europe (EU, 2007) and facilitated by intergovernmental river basin organizations like the International Commission for the Protection of the Rhine (ICPR) or the International Commission for the Protection of the Danube River (ICPDR). In these organizations, the member countries need to jointly evaluate the effectiveness of various risk reducing-strategies over time (ICPR, 2006, 2007). However, different damage modelling approaches are applied by riparian countries (Meyer and Messner, 2005), and these can yield substantial differences in terms of absolute damage values (de Moel and Aerts, 2011), and the results are often difficult to compare. Therefore, usually one flood damage modelling approach needs to be chosen and agreed upon by the various stakeholders when conducting such joint assessments (ICPR, 2001a; Silva and Reuter, 2006). This makes it important to understand to what extent various damage modelling approaches used in different riparian countries or regions influence estimates of relative changes in flood damage developments. In order to improve such relative estimates, it is especially of interest to understand whether variations between the damage modelling approaches stem from the uncertainties in the damage curves or from differences in estimating the exposed asset values linked to these curves.

Studies that have evaluated the possible range of flood damage developments have commonly addressed the uncertainties originating from projections of socio-economic development and climate change by applying alternative scenarios for land-use change and flood probabilities linked to climate change (Hall et al., 2005; ABI, 2009; Bouwer et al., 2010). In addition, when focusing on the river Rhine in North-Western Europe, several scenario studies have evaluated the relative influence of climate and land-use change on flood damage developments (Klijn et al., 2007; Aerts et al., 2008; te Linde et al., 2011). However, it is largely unknown how different flood damage modelling approaches influence estimates of relative changes in the development of potential flood damage, as compared with those scenarios. This is because, although previous studies have applied a range of scenarios, they have generally only used one damage modelling approach to assess changes in flood damage.

The aim of the current study is, therefore, to evaluate the reliability of relative estimates of flood damage developments for the river Rhine with regard to different flood damage modelling approaches. We explain the differences for two damage models through simulating the effects of socio-economic development, reflected by land-use scenarios, on flood damage between 1990 and 2030. These two damage models have been used earlier to assess potential flood damage in the Rhine Basin: the flood damage model used for the development of the Rhine Atlas (ICPR, 2001a) and the one called the “Damage Scanner” (Klijn et al., 2007). Furthermore, we will investigate whether variations in absolute and relative damage estimates between the approaches result from differences in damage functions or from the estimation of the exposed asset values. Finally, the variations stemming from the application of different flood damage modelling approaches are compared with the uncertainties originating from land-use projections.

The remainder of the article proceeds as follows: Sect. 2 provides a general introduction to flood damage modelling approaches. Section 3 describes the study area and discusses the data and methods applied in the present study. The results are presented and discussed in Sect. 4. Section 5 concludes.

2 Flood damage assessments

There is a wide range of approaches to estimate potential flood damage, mostly depending on the spatial and temporal scale of the analysis as well as on its purpose and the availability of data and resources (Messner et al., 2007). Micro-scale assessments are characterized by a high level of detail, and, from the exposure side, they usually take single objects into account that are at risk from flooding, such as buildings, vehicles or infrastructure. From the hazard side, consideration is given to location specific flood characteristics, such as flow velocity, water depth, or contamination of flood water. In meso-scale flood damage analyses, as applied in the
present study, exposure information is usually based on the spatial aggregation of the exposed assets. Most commonly, meso-scale models combine information on land-use types for estimating assets at risk and information on inundation height for simulating the hazard. On the macro-scale, even larger units, such as municipalities, regions or countries, are used for damage estimations (Merz et al., 2010).

A standard approach to quantify flood damage is the use of stage-damage functions (Smith, 1994; Merz et al., 2007). These can be developed using two different methodological approaches (Merz et al., 2010). First, empirical damage functions can be constructed on the basis of observed flood damage data. For example, the HOWAS database, which was maintained in Germany (Merz et al., 2004) and is now integrated in the HOWAS21 database\(^1\), is a collection of observed damage cases, and has provided the basis for the derivation of the damage functions applied in the Rhine Atlas (ICPR, 2001a). Second, stage-damage functions can be derived using synthetic approaches, which are based on expert judgement. Following this approach, experts from, for instance the insurance industry, estimate the damage that can be expected for different types of objects and assets when certain flood characteristics occur. The HIS-SSM damage model, which is the standard software in the Netherlands to evaluate flood damage (Kok et al., 2005), and the Multicoloured Manual in the UK (Penning-Rosell et al., 2005) are both examples of this approach, although the HIS-SSM model also includes some empirical information. For more detailed overviews of flood damage assessment approaches and their various aspects, see Merz et al. (2010) and Messner et al. (2007).

3 Study area, data, and methods

Our research approach to study the reliability of estimates of relative changes in the development of potential flood damage is shown in Fig. 1. The river Rhine was chosen as a case study because it is the largest and economically most important river in Western Europe. Furthermore, the river basin is shared by nine countries, which need to engage in transboundary flood risk management and, being part of this, need to jointly evaluate the effectiveness of risk-reducing measures (ICPR, 2007). We apply two meso-scale flood damage models that have been commonly used for flood damage assessments along the Rhine: the Rhine Atlas model used by the ICPR (ICPR, 2001a), and what is called the “Damage Scanner” (Klijn et al., 2007). The two models have been selected because they show a large variability in absolute damage estimates (de Moel and Aerts, 2011), and have been derived using different methodological approaches. In both models, potential flood damage is quantified with the help of depth-damage functions that define for each land-use type the potential damage, depending on the given inundation-depth at the respective location. Other flood characteristics are not considered. The absolute flood damage estimates derived from applying the two models in the case study area are used to evaluate the relative change (in percentages) in flood damage owing to land-use change between 1990 and 2000, as well as between 2000 and the two land-use projections for 2030. Estimates of relative changes according to the two models are compared with each other in order to gain insights into the reliability of relative estimates of flood damage developments with regard to different flood damage modelling approaches. The remainder of this section describes the study area, the data, and the methods applied in the present study in greater detail.

3.1 River Rhine basin

The river Rhine originates in the Swiss Alps and flows through Austria, Switzerland, France and Germany into the Netherlands, where it eventually bifurcates into the North Sea and the Lake IJssel. The river basin area is about 197 000 km\(^2\), and is shared by nine countries. The basin is densely populated, and, in total, about 58 million people live in the area (ICPR, 2001a, 2008). Especially since the 19th century, the Rhine has developed into an important traffic route and is today one of the world’s most trafficked and used waterways. It connects one of the world’s largest sea harbours in Rotterdam, the Netherlands, with the world’s largest inland harbour in Duisburg and other industrial complexes in Germany, France, and Switzerland. Every year, about 200 000 vessels cross the Dutch-German border, transporting approximately 200 million tons of goods (ICPR, 2008).

To aid shipping and industrialization, the main river channel has undergone substantial changes and has been rectified and canalized. From the original 8000 km\(^2\) of floodplains, less than 15 percent remain (ICPR, 2008). Moreover, the canalization and rectification of the riverbed has caused an acceleration of flood wave propagation in the Rhine canal (Lammersen et al., 2002). These developments have led to an increasing risk of flooding, and the ICPR estimates that about 10 million people live at risk from extreme flooding (ICPR, 2001a). Flood protection levels, expressed as average return periods of peak discharges that can be withstood by dykes and embankments, vary along the river and range from 1/200 per year in the Upper Rhine and parts of the middle Rhine, 1/500 per year in the lower Rhine, to 1/1250 and 1/2000 per year in the Netherlands. Two major floods occurred in 1993 and 1995 along the Rhine that caused substantial economic damage in Germany (Kron and Thumerer, 2002) and the preventive evacuation of about 250 000 people in the Netherlands.

\(^1\) http://nadine-ws.gfz-potsdam.de:8080/howasPortal/client/start
3.2 Inundation data

For the Rhine and its floodplains up to the Netherlands, we use the inundation map that was developed by the International Commission for the Protection of the Rhine (ICPR, 2001a). A 100 × 100 m grid map was derived from the original vector data set that provides information on the flood extent and seven different classes of inundation depths, ranging from 50 cm to 5 m and above. For the part of the Rhine in the Netherlands, updated and improved information on potentially inundated areas is available in the form of the Dutch “Risicokaart”. The “Risicokaart”, which is based on hydro-dynamical modelling, shows the maximum inundation depths that are derived from numerous simulations of dyke breaches (Wouters and Holterman, 2007). It was originally provided at a spatial resolution of 50 m × 50 m and was resampled to a 100 m × 100 m resolution grid. The combined inundation map for the Rhine channel shows flood extent and depths for extreme discharge events, with events ranging from an average frequency of occurrence of 1/200 yr at the Upper Rhine to 1/2000 yr in the Dutch delta area, and has been applied earlier in a study on future flood damage (e.g. te Linde et al., 2011). Only those areas in the Dutch delta that are prone to river flooding from the Rhine were included. Those areas that face flood risk, mainly from coastal storm surges in combination with high but not extreme river discharges, were not included. It should be noted that the inundation map used in the present study does not represent a realistic flooding scenario, because there will never be enough water in a single event to inundate the entire area. Instead, it provides an indication of the potential maximum hazard facing all the economic assets exposed to floods along the Rhine.

3.3 Land-use information

Information on land use is based on CORINE land cover grids for the years 1990 and 2000 (Bossard et al., 2000), which are available at a spatial resolution of 100 m × 100 m from the European Environment Agency. The grids were aggregated to the six and nine land-use classes of the two damage models, respectively (see Table 1). Two land use projections for 2030 were derived from a land use model known as the “Land Use Scanner” (Hilferink and Rietveld, 1999; Looonen and Koomen, 2009), which was adapted for the Rhine basin (te Linde et al., 2011). The “Land Use Scanner” model has been applied in numerous research projects and policy appraisals in the Netherlands and other European countries, including assessments of future flood risk (Borsboom-van Beurden et al., 2007; Aerts et al., 2008; Maaskant et al., 2009; Bouwer et al., 2010; te Linde et al., 2011). The model
Table 1. Maximum damage values for land-use classes of the Rhine Atlas and the Damage Scanner models (at 2000 prices).

<table>
<thead>
<tr>
<th>Rhine Atlas</th>
<th>Max. damage (€/m²)</th>
<th>Damage Scanner</th>
<th>Max. damage (€/m²)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Residential</td>
<td>233</td>
<td>Residential – high density</td>
<td>910</td>
</tr>
<tr>
<td>Residential mobile</td>
<td>55</td>
<td>Residential – low density</td>
<td>400</td>
</tr>
<tr>
<td>Industry</td>
<td>246</td>
<td>Commercial</td>
<td>600</td>
</tr>
<tr>
<td>Industry mobile</td>
<td>82</td>
<td>Infrastructure</td>
<td>190</td>
</tr>
<tr>
<td>Infrastructure</td>
<td>250</td>
<td>Agriculture</td>
<td>20</td>
</tr>
<tr>
<td>Infrastructure mobile</td>
<td>2</td>
<td>Pasture</td>
<td>10</td>
</tr>
<tr>
<td>Agriculture</td>
<td>7</td>
<td>Nature/Forest</td>
<td>20</td>
</tr>
<tr>
<td>Forest</td>
<td>1</td>
<td>Building lot</td>
<td>130</td>
</tr>
<tr>
<td>Other</td>
<td>0</td>
<td>Recreation</td>
<td>30</td>
</tr>
</tbody>
</table>

is based on microeconomic theory, and uses socio-economic scenarios to simulate land-use patterns on the basis of spatial claims (expected quantity of land-use change), suitability, and policy maps (Fig. 2).

The existing Land Use Scanner model for the Rhine (see te Linde et al., 2011) has been forced with two contrasting socio-economic scenarios to produce future land-use maps (Fig. 2). These scenarios are derived from the EURURALIS project (Verburg et al., 2008; Verburg and Overmars, 2009), and describe future spatial claims for the Rhine area. The Global Economy scenario (GE) assumes a world with high economic and population growth, international economic integration and a strong influence of private interests. A weak government is assumed that enforces little environmental regulation. The GE scenario results in a land-use projection that sees a large increase in urban land use. No restrictions in terms of urban development are applied to areas at risk of flooding due to the weak role of the government assumed for this scenario. In contrast, the Regional Communities scenario (RC) assumes a world with little economic and population growth and a strong regional focus. In this scenario, a strong government is foreseen that enforces strict environmental regulations such as spatial zoning in flood-prone areas. The RC scenario leads to a land-use change projection that sees far less urban development, which is considerably restricted in areas at risk from flooding. The GE and RC scenarios are comparable to the A1 and B2 scenarios of the Intergovernmental Panel on Climate Change, respectively (IPCC, 2000). While no probability is attached to each of these two scenarios, we assume that the two scenarios reflect the possible bandwidth of future changes in land use and urban development. For a more detailed description on the downscaling approach and the development of these land-use projections, the reader is referred to te Linde et al. (2011).

3.4 Damage models

On the basis of the flood inundation map (Sect. 3.2) and the four land-use maps (Sect. 3.3), two flood damage modelling approaches are used to calculate potential flood damage along the Rhine for the years 1990, 2000, and the two projections for 2030.

3.4.1 Rhine Atlas damage model

The Rhine Atlas damage model (RAM) was developed for the International Commission for the Protection of the Rhine (ICPR, 2001b). The method uses five depth-damage functions to calculate potential flood damage. The functions were constructed on the basis of the HOWAS database, and are thus empirical in nature, even though additional expert judgement has been used to select and combine entries from the HOWAS database (ICPR, 2001b). Even though the RAM has been developed to estimate potential flood damage along the trans-boundary river Rhine, it is thus largely based on empirical damage data from Germany.

The depth-damage functions are linked to five re-classified CORINE land cover classes, of which three reflect urban land-use types (residential, industrial, and infrastructure). For the three urban categories, separate damage values for property and contents are used in the RAM (see Table 1) (ICPR, 2001a). The model considers direct tangible flood damage that is caused by physical contact between floodwater and the economic assets at risk. Not reflected in the RAM is indirect damage, such as losses due to the disruption of production processes, and also intangible damage, such as loss of life.

3.4.2 Damage Scanner model

The Damage Scanner model (DSM) also calculates potential flood damage by linking information on land use and
information on inundation depth with seven depth-damage functions (see Table 1 and Fig. 3). It has been used in various studies that assess the impact of future land use on potential flood damage at various scales: amongst others, along the Rhine and especially in the Rhine delta (Aerts et al., 2008; Bouwer et al., 2010; te Linde et al., 2011; de Moel et al., 2011). As the functions of the DSM are based on the HIS-SSM, which is the standard software in the Netherlands to assess flood damage, they are based on some empirical information as well as expert judgment and thus are mainly synthetic in nature. The model predominantly estimates direct tangible damage. However, the DSM also implicitly includes a share of 5 per cent indirect damage. This latter type of damage reflects potential losses in economic turnover due to traffic interruption or emergency costs.

3.4.3 Comparison of the RAM versus the DSM

Differences between the two models can be found both in the applied damage curves and in the maximum damage values linked to the respective damage categories. Figure 3 displays the shape of the damage curves of the RAM and the DSM. These curves are used to calculate the fraction of the maximum damage (damage factor) occurring in a grid cell, based on the respective inundation depth.

When comparing the residential damage curves from both models (solid red lines), it becomes apparent that the curve of the DSM is steeper compared with that of the RAM. For instance, at 4 m water depth, the RAM curve gives a damage factor of about 0.4, while the DSM curve gives a damage factor of about 0.8. In addition, both models differ in their assumption at which water level the maximum damage is reached. According to the DSM, maximum damage of all land-use types is reached at a water level of about 5 m. In contrast, the RAM assumes that a water level of 5 m results in about 60 per cent of the maximum damage for residential and commercial areas. These differences in the functions reflect the uncertainties still associated with depth-damage curves in flood damage modelling (e.g. Freni et al., 2010). These uncertainties stem from the large variability of observed damage even among similar elements at risk. Two buildings that share the same physical characteristics and that are situated next to each other can nevertheless experience largely different amounts of damage during the same flood event. This is because not only flood impact parameters, such as flow velocity and the contamination of flood water, but also flood resistance parameters, such as flood experience and precautionary measures, can significantly differ even within short distances (Merz et al., 2010).

Moreover, both models differ substantially in terms of the maximum damage values that are linked to the damage categories, represented by the respective land-use types. An overview of the maximum damage values at 2000 price levels for respective damage categories according to both models is provided in Table 1.
Table 1 shows that the maximum damage for similar damage categories is significantly higher in the DSM compared with the RAM. For instance, the residential category of the RAM has a maximum damage of 288 € m$^{-2}$ (sum of residential and residential mobile), while the maximum damage of the DSM is 910 € m$^{-2}$ for high density urban areas, and 400 € m$^{-2}$ for low density urban areas. As a comparison, the standard flood loss model in the UK estimates a maximum damage value of 750 £ m$^{-2}$ (865 € m$^{-2}$)\(^2\) for a residential home (Penning-Rowsell et al., 2005). Likewise, the industrial damage category also shows a marked difference between the RAM and the DSM: it has a maximum damage of 328 € m$^{-2}$ in the RAM compared with 600 € m$^{-2}$ in the DSM. While these urban land-use classes have a maximum damage that differs roughly by a factor of 2, the difference between both agricultural land use and nature areas is even larger, by a factor of 2.8 and 20, respectively.

There are a number of reasons for the large differences in terms of maximum damage values between the RAM and the DSM. While the RAM uses depreciated asset values, the DSM is, at least in some instances, based on replacement costs, such as those for building structures and contents (Briene et al., 2002). Differences in price levels among the riparian countries cannot serve as an explanation for the large variations in terms of maximum damage values, as purchasing power is only marginally different (see ICPR, 2001b). Damage to vehicles, which can make a significant contribution to total damage of 2 per cent to 7 per cent (ICPR, 2001a), is not considered in the maximum damage values of the RAM (ICPR, 2001a), while replacement values for vehicles are included in the DSM (Briene et al., 2002). The explanation for the very large difference in the maximum damage of agricultural areas between the two models is that agricultural practices in the Netherlands are more capital intensive (ICPR, 2001a). Additionally, the DSM takes into account that, in grid cells with an agricultural land use, there are also buildings with inventory present and not just crops. Furthermore, the DSM also implicitly comprises approximately 5 per cent of indirect damage as a surcharge on direct damage. Indirect damage, which can make up a substantial share of total flood damage, refers to losses that are induced by the flood event itself but occur with a time delay or outside the flooded area. Hallegatte (2008), for instance, published a study on the relation between direct and indirect damage. He investigates the response of the regional economy of Louisiana to the landfall of hurricane Katrina using an input-output model. The simulation results show that disturbed economic processes aggravate direct damage. While the direct damage of hurricane Katrina is estimated at $107 bn, the indirect effects are estimated at an additional $42 bn, which is hence 28 per cent of total losses (Hallegatte, 2008).

\(^2\)According to exchange rates of 19 August 2011

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Fig. 3. Damage curves for the Rhine Atlas and the Damage Scanner model (Adapted from de Moel and Aerts, 2011).
4 Results and discussion

4.1 Absolute flood damage estimates

An overview of potential damage along the Rhine according to the RAM and the DSM model for 1990, 2000, and to the two socio-economic scenarios (2030), as well as their difference factors, is provided in Table 2. Both model results differ significantly in terms of absolute damage estimates by a factor ranging from 3.5 to 3.8. In general, the RAM gives much lower results compared with the DSM. The results are in line with several other studies that have stressed the variations related to the assessment of direct flood damage (e.g., Merz et al., 2004; Apel et al., 2008; Freni et al., 2010). De Moel and Aerts (2011), for instance, estimate flood damage for a Dutch polder area using three different damage modelling approaches, and find differences up to a factor of 4.

The large variations in absolute damage estimates can be attributed to existing uncertainties in the shapes of damage curves and to methodological differences for estimating the underlying exposed asset values of the respective damage categories, which were discussed in Sect. 3.4.3.

To identify whether variations between the models in absolute damage estimates result from differences in the underlying maximum damage values or from differences in the damage curves, we performed an additional model-run with the RAM, using the maximum damage values of the DSM. This model-run is referred to as RAM\textsuperscript{DSM}. By comparing the results of the RAM\textsuperscript{DSM} with the results of the RAM, we can estimate how the differences in maximum damage values influence absolute damage estimates, because both the RAM and the RAM\textsuperscript{DSM} use the same damage functions. By comparing the results of the DSM with the results of the RAM\textsuperscript{DSM}, we can estimate how the differences in damage functions influence the absolute damage estimates, because both the DSM and the RAM\textsuperscript{DSM} use the same maximum damage values. An overview of the absolute damage estimates for the RAM\textsuperscript{DSM} and the difference factors between the RAM and the RAM\textsuperscript{DSM}, and between the DSM and the RAM\textsuperscript{DSM}, are provided in Table 3. It is shown that differences in the maximum damage values have a smaller influence on variations in absolute damage estimates between the two models (by a factor 1.8 for all time steps) than differences in the functions (by a factor of 1.93–2.13).

Other studies that have investigated the influence of different damage functions on absolute damage estimates have come to similar conclusions. For instance, Apel et al. (2009) compare the results of several damage modelling approaches with recorded flood damage and find that differences in damage curves can lead to relative errors ranging from −87 per cent to 34 per cent. Since the same exposure data was used for all model runs, the influence of different maximum damage values on absolute damage estimates was not assessed in the study of Apel et al. (2009).

4.2 Estimates of relative changes in the development of potential flood damage

The large differences in absolute damage estimates that result from the application of the two different damage modelling approaches is the reason for evaluating the reliability of relative estimates. We therefore assess the results from the two models in terms of relative changes in potential flood damage for the periods 1990 to 2000 and 2000 to 2030 (using the GE and RC scenarios). As can be seen from Fig. 3, the RAM model estimates consistently higher relative increases in potential flood damage compared with the DSM. While the RAM model foresees a relative increase of 39.1 per cent for the GE scenario and 11.8 per cent for the RC scenario, the DSM projects an increase of 27.3 per cent and 8.4 per cent, respectively. For all time steps, the RAM model gives relative estimates, which are about 1.1 to 1.4 times higher than the DSM model. It should be noted that this variation is considerably smaller compared with differences observed for the absolute damage estimates of the two models (Table 2). This finding is in line with previous studies, which showed that estimates of proportional changes in flood damage are much more robust compared with absolute estimates (de Moel and Aerts, 2011). Moreover, differences between the two models when estimating relative changes are considerably smaller compared with uncertainties inherent in future projections of socio-economic development, as reflected by the two contrasting land-use scenarios. With respect to potential damage, the two land-use projections differ by more than a factor of 3 for both models (Fig. 4). To identify whether variations between the models in terms of relative estimates result...
from differences in the underlying maximum damage values or from differences in the damage curves, we again compare the RAM\textsuperscript{DSM} with the RAM and with the DSM. An overview of the relative estimates of flood damage developments of the RAM\textsuperscript{DSM} and the difference factor between the relative estimates of the RAM\textsuperscript{DSM} and RAM, and between the relative estimates of the RAM\textsuperscript{DSM} and the DSM are provided in Table 4. It is shown that differences in the relative estimates of flood damage developments are predominantly influenced by differences in the damage functions between the two models. The RAM and the RAM\textsuperscript{DSM}, which use the same damage functions but different maximum damage values, perform almost identically (by a factor of 0.99 to 1.01). In contrast, the comparison between the RAM\textsuperscript{DSM} and the DSM, which use the same maximum damage values but different functions, show difference factors of 1.15 to 1.42.

The fact that differences in maximum damage values have hardly an influence on relative estimates of flood damage developments in the present study can be explained by the proportion of grid cells that change in land use, and grid cells that remain unchanged between two time steps. For instance, when looking at the GE scenario, only 2.5 per cent of all grid cells change in land use between 2000 and 2030 in the flood prone area. For the majority of grid cells that do not change in land use, differences in maximum damage values between the two models have no influence on relative estimates of flood damage developments. This is because, even though the application of the RAM and the RAM\textsuperscript{DSM} leads to different absolute damage values due to the differences in maximum damage values, both models foresee no change in damage between 2000 and 2030 GE for a grid cell that does not change in land use. In contrast, differences in the damage functions between the two models lead to changes in relative estimates for all grid cells. Because only a small share of grid cells change in land use, the influence of differences in maximum damage values on relative estimates of changes in the development of flood damage is small.

**Table 4.** Relative estimate of damage developments of the model-run RAM\textsuperscript{DSM} and difference factors between relative estimates of the RAM and the RAM\textsuperscript{DSM}, and between the RAM\textsuperscript{DSM} and the DSM.

<table>
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</thead>
<tbody>
<tr>
<td>RAM\textsuperscript{DSM} Diff. due to max. damage values*</td>
<td>4.16 %</td>
<td>11.96 %</td>
<td>38.85 %</td>
</tr>
<tr>
<td>RAM\textsuperscript{DSM} Diff. due to function**</td>
<td>0.99</td>
<td>1.01</td>
<td>0.99</td>
</tr>
<tr>
<td>RAM\textsuperscript{DSM} to RAM (difference factor)</td>
<td>1.15</td>
<td>1.41</td>
<td>1.42</td>
</tr>
<tr>
<td>RAM\textsuperscript{DSM} to DSM (difference factor)</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

A different result is obtained when focusing only on grid cells that change in land use. For those grid cells, the influence of difference in maximum damage values is substantial, and can be explained by the different weighting of respective damage categories within the individual models. Table 5 shows the percentage that each damage category, represented by land-use classes, contributes to total damage in 2000 for both models. It is shown that urban land-use classes (residential and commercial) contribute about 98 per cent of the total damage in the RAM model. In the DSM model, this percentage is considerably lower (79 per cent), and agricultural land use also contributes significantly to the overall damage (14.8 per cent).

The varying contributions of the respective damage categories to the total damage of the two models is a result of the different ratio of maximum damage values across the assigned damage categories within each of the two models (see Table 1). For example, the difference between the maximum damage of residential and agricultural land use is about a factor of 41 (288 € m\textsuperscript{−2} divided by 7 € m\textsuperscript{−2}) in the RAM model, while the difference between low-density residential areas, which comprise most of the residential areas, and agriculture differs only by a factor of 20 (400 € m\textsuperscript{−2} divided by 20 € m\textsuperscript{−2}) in the DSM model. Because these ratios are so different, a change from agriculture to low-density residential land use in the flood-prone area has a different effect on the increase in potential flood damage between the two models. This will be further illustrated below by looking at the land-use change and the corresponding change in flood damage between 2000 and 2030 (GE scenario).

The GE scenario represents a “strong growth scenario”, which results in a considerable increase in urban land-use types, mainly at the expense of agricultural areas. Table 6 provides a matrix of net changes among all land-use classes between 2000 and 2030 (GE), and insights into the specific changes that occur among land-use types between 2000 and 2030 GE. The fact that net changes are presented means that conversions between two land-use types in both directions are considered. It shows that predominantly agricultural areas change into low-density residential (32 328 hectares) and commercial areas (6504 hectares). The increase in
Table 5. Relative share of respective land-use types to total damage (2000) for the Rhine Atlas and the Damage Scanner models.

<table>
<thead>
<tr>
<th>Land use</th>
<th>Rhine Atlas Total damage (€)</th>
<th>Rhine Atlas Total damage (%)</th>
<th>Damage Scanner Total damage (€)</th>
<th>Damage Scanner Total damage (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Residential</td>
<td>58 248</td>
<td>74.9</td>
<td>187 707</td>
<td>62.5</td>
</tr>
<tr>
<td>Commercial</td>
<td>17 919</td>
<td>23.0</td>
<td>49 359</td>
<td>16.4</td>
</tr>
<tr>
<td>Infrastructure</td>
<td>1231</td>
<td>1.6</td>
<td>6339</td>
<td>2.1</td>
</tr>
<tr>
<td>Nature/Forestry</td>
<td>6</td>
<td>0.0</td>
<td>7626</td>
<td>2.5</td>
</tr>
<tr>
<td>Agriculture</td>
<td>346</td>
<td>0.4</td>
<td>44 487</td>
<td>14.8</td>
</tr>
<tr>
<td>Construction/Mines</td>
<td>–</td>
<td>–</td>
<td>2840</td>
<td>0.9</td>
</tr>
<tr>
<td>Recreation</td>
<td>–</td>
<td>–</td>
<td>2104</td>
<td>0.7</td>
</tr>
<tr>
<td>Total</td>
<td>77 750</td>
<td>100</td>
<td>300 463</td>
<td>100</td>
</tr>
</tbody>
</table>

high-density urban areas is mainly due to the conversion of formerly low-density residential areas (Table 6), and thus reflects an increase in urban density. Important changes among land-use types that lead to large changes in potential flood damage are highlighted in colour.

These transformations from mainly agricultural areas to urban land-use types lead to a large increase in potential flood damage for these grid cells, given the importance of urban areas in both models. However, differences exist between the two models, because in the RAM model, urban damage categories make up a larger share of total damage compared with the DSM model (see Table 5). This also leads to a larger increase in potential damage for grid cells that change in land use according to the RAM over time compared with the DSM.

To demonstrate this in greater detail, only those grid cells that change between agriculture and low-density residential areas between 2000 and 2030 GE are extracted and analysed separately with respect to absolute and relative changes in potential flood damage. The analysis shows that the shift of the 32 328 hectares from agricultural to low-density residential areas results in a damage increase for these grid cells of 478 per cent in the RAM model, while it leads to an increase of 341 per cent according to the DSM model (Table 7). In addition to the predominant change between agricultural and low-density residential areas, other important land-use changes with a large effect on potential damage were analysed the same way. An overview of the specific effect of important changes among land-use types in terms of absolute and relative changes in potential damage for the two models is provided in Table 7.

Insights into the question: “which of the two models provides more reliable estimates of flood damage developments for grid cells that change in land use?”, could be provided by information on how different damage categories have contributed to overall flood damage for observed flood events along the Rhine. Recorded damage data of an extreme event would be of interest for the present study, because inundation maps of potential extreme flood events are used as input for the two damage models. However, according to our knowledge, such detailed data that breaks down recorded flood damage into different damage categories is not available for the Rhine. This also holds true in general: while the hydrological parameters of a flood are commonly evaluated in the aftermath of an event (e.g. Landesamt für Umwelt Baden-Württemberg, 2000), comprehensive data on observed damage following from these events is scarce (e.g. Sächsische Staatskanzlei, 2003).

A major flood event, for which more detailed damage data could be obtained, occurred along the Elbe in 2002 and predominantly affected the federal state of Saxony (Sächsische Staatskanzlei, 2003). Table 8 shows overall recorded damage for this event, and also provides the contribution of several damage categories to overall losses. Even though these data are not necessarily comparable to the Rhine due to geographical differences, two aspects also seem to provide valid insights for the present study. First, the data of the Elbe flood strongly suggest that both models remarkably underestimate damage to infrastructure. According to the RAM damage model and the DSM model, infrastructure does not significantly contribute to overall losses, with a share of 1.6 per cent and 2.1 per cent, respectively (see Table 5). However, during the Elbe flood, damage to municipal infrastructure alone comprised about 20 per cent of overall losses. Within this damage category, damage to roads and bridges (36 per cent) and damage to water protection and flood defences (11 per cent) were the largest items. Also, with regard to state-run infrastructure, which made up 15 per cent of overall reported losses, damage to roads and bridges, as well as to water protection and flood defences, were the two largest items with 12 and 63 per cent, respectively. The fact that damage to roads and bridges is hardly reflected in meso-scale flood damage assessments, as also applied in the present study, can be explained by the resolution of the land-use maps that are usually employed in such studies. Often, the latter have a resolution of 100 m × 100 m or coarser. Because they have
Table 6. Matrix of net changes (in hectares) among the respective land use types between 2000 and 2030 GE.

<table>
<thead>
<tr>
<th></th>
<th>Residential HD</th>
<th>Residential LD</th>
<th>Commercial</th>
<th>Infrastructure</th>
<th>Construction</th>
<th>Recreation</th>
<th>Nature</th>
<th>Agriculture</th>
</tr>
</thead>
<tbody>
<tr>
<td>Residential HD</td>
<td>0</td>
<td>10499</td>
<td>0</td>
<td>−1398</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Residential LD</td>
<td></td>
<td>−599</td>
<td>0</td>
<td>−2</td>
<td>−20</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Commercial</td>
<td>−26</td>
<td>−23</td>
<td>22</td>
<td>−26</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Infrastructure</td>
<td></td>
<td>0</td>
<td>0</td>
<td>−2</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Construction</td>
<td></td>
<td></td>
<td>0</td>
<td>−282</td>
<td>1410</td>
<td>369</td>
<td>13</td>
<td>0</td>
</tr>
<tr>
<td>Recreation</td>
<td></td>
<td>14777</td>
<td>1071</td>
<td>13</td>
<td>17</td>
<td>669</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Nature</td>
<td>−26</td>
<td>4776</td>
<td>1071</td>
<td>13</td>
<td>17</td>
<td>669</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Agriculture</td>
<td>1253</td>
<td>32328</td>
<td>6504</td>
<td>136</td>
<td>125</td>
<td>3464</td>
<td>10985</td>
<td>0</td>
</tr>
<tr>
<td>Total Increase (ha)</td>
<td>10026</td>
<td>27437</td>
<td>9943</td>
<td>210</td>
<td>91</td>
<td>2623</td>
<td>4465</td>
<td>−54795</td>
</tr>
</tbody>
</table>

Note: the specific effect of the highlighted changes among land-use types in terms of changes in potential damage is provided in Table 7.

Table 7. Specific effect of important changes among land-use types in terms of potential flood damage for the two models.

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>2000</td>
<td>2030 GE</td>
</tr>
<tr>
<td>Agriculture and Residential (LD)</td>
<td>962</td>
<td>4242</td>
</tr>
<tr>
<td>Residential (LD) and Residential (HD)</td>
<td>1831</td>
<td>3270</td>
</tr>
<tr>
<td>Nature and Residential (LD)</td>
<td>30</td>
<td>460</td>
</tr>
<tr>
<td>Agriculture and Commercial</td>
<td>139</td>
<td>1316</td>
</tr>
</tbody>
</table>

Table 8. Recorded flood damages in Saxony during the Elbe flood in 2002

<table>
<thead>
<tr>
<th>Observed damage</th>
<th>Absolute (Mill. €)</th>
<th>% of total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Residential Buildings</td>
<td>1706</td>
<td>27.5</td>
</tr>
<tr>
<td>Building contents</td>
<td>529</td>
<td>8.5</td>
</tr>
<tr>
<td>Industry</td>
<td>1420</td>
<td>22.9</td>
</tr>
<tr>
<td>Municipal Infrastructure</td>
<td>1287</td>
<td>20.8</td>
</tr>
<tr>
<td>state-run infrastructure</td>
<td>928</td>
<td>15.0</td>
</tr>
<tr>
<td>Costs of disaster operation</td>
<td>136</td>
<td>2.2</td>
</tr>
<tr>
<td>Cultural facilities</td>
<td>111</td>
<td>1.8</td>
</tr>
<tr>
<td>Agriculture and Forestry</td>
<td>79</td>
<td>1.3</td>
</tr>
<tr>
<td>Total</td>
<td>6196</td>
<td>100.0</td>
</tr>
</tbody>
</table>

Source: adapted from Sächsische Staatskanzlei (2003).

such their narrow shapes, infrastructural elements are underrepresented in land-use maps of this resolution. Bouwer et al. (2009) show that the share of potential flood damage to infrastructure in overall losses increases from 3.9 to 8 per cent in a Dutch polder area, when switching from a 100 m × 100 m to a 25 m × 25 m grid resolution. Second, the recorded damage data of the Elbe flood suggest that both models might overestimate the contribution of residential areas to overall losses. According to the RAM and the DSM, the residential damage category contributes 75 per cent and 62 per cent, respectively (see Table 5). However, during the Elbe flood, damage to residential buildings and their contents comprised 36 per cent of overall losses. Since these observations are based on a single event only, it should be noted that, in order to draw more general conclusions, more data on recorded flood damage and the contribution of different damage categories to overall losses is necessary.

5 Conclusions

The present study has investigated the reliability of estimates of relative changes in the development of potential flood damage with regard to the uncertainties stemming from flood damage modelling approaches. We find that two flood damage modelling approaches differ by a factor of 3.5 to 3.8 for absolute damage estimates. These differences are attributed to both uncertainties in the depth-damage curves and methodological differences for estimating the underlying maximum damage values of the respective damage categories. Variations in maximum damage values between the two models are found to have a smaller influence (by a factor of 1.8) than variations in the damage curves (by a factor of 1.9–2.1). The variations in maximum damage values stem from different methodological approaches, and from the choice of economic parameters, such as depreciated asset values versus replacement cost, mean insured value or standardized constructions costs, which are used to derive the...
asset values linked to the damage curves. As this can lead to large differences in terms of absolute damage estimates, it should always be clearly specified which type of damage and assets are taken into account, and how the attached asset values were derived.

With respect to relative estimates of flood damage developments, it was found that, when estimating relative flood damage developments, both modelling approaches provide very similar results that differ by about a factor of 1.4. This number is considerably lower than it is for the difference in absolute estimates between the models, and also smaller than it is for the uncertainty inherent in projections of future exposure, as represented by contrasting land-use projections. With respect to potential damage, the two land-use projections applied in the present study result in differences in damage of more than a factor of 3. In contrast to variations in absolute damage estimates, the differences between the damage modelling approaches in terms of relative changes can be attributed predominantly to the differences in the damage functions between the models. This is because the differences in maximum damage values only have an influence on grid cells that change in land use. As only a relatively small share of the grid cells changes between the two time steps, e.g. 2.5 per cent of all grid cells change between 2000 and 2030 GE, the differences in maximum damage values between the two models have hardly any influence on relative estimates of flood damage developments. The differences in damage functions, in contrast, apply to all grid cells in the flood-prone area, which explains the larger influence on variations in relative estimates, with difference factors ranging from 1.1 to 1.4.

These findings provide valuable insights for intergovernmental river-basin organizations like the ICPR. Such organizations are increasingly required to engage in trans-boundary flood risk assessments under the EU Water Framework Directive, and usually need to choose among various flood damage modelling approaches when doing so. We show that the influence of different modelling approaches on estimates of relative changes in the development of flood damage is small compared with other sources of uncertainties encountered in flood damage projections.

In order to improve the reliability of relative estimates of flood damage developments, future research should focus on reducing the uncertainties of stage-damage functions, which originates from the huge variability of observed damage even among similar elements at risk. Although it is only possible to integrate this enormous variability in flood damage modelling approaches to a limited extent, it has been shown that model performance can be improved by integrating several damage-influencing parameters in multi-parameter damage modelling (e.g. Elmer et al., 2010). As this requires detailed data on damage processes at the level of individual objects, data collection after flood events remains a crucial activity.

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