1 Regionalisation of population growth projections in coastal exposure analysis

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12 Abstract

13 Large-area coastal exposure and impact analysis has focussed on using sea-level rise (SLR) scenarios and has 14 placed little emphasis on socioeconomic scenarios, while neglecting spatial variations of population dynamics. We 15 use the Dynamic Interactive Vulnerability Assessment (DIVA) Framework to assess the population exposed to 1 16 in 100-year coastal flood events under different population scenarios, that are consistent with the Shared 17 Socioeconomic Pathways (SSPs); and different SLR scenarios, derived from the Representative Concentration 18 Pathways (RCPs); and analyse the effect of accounting for regionalised population dynamics on population 19 exposure until 2100. In a reference approach, we use homogeneous population growth on national level. In the 20 regionalisation approaches, we test existing spatially explicit projections that also account for urbanisation, coastal 21 migration and urban sprawl. Our results show that projected global exposure in 2100 ranges from 100 million to 22 260 million, depending on the combination of SLR and population scenarios and method used for regionalising 23 the population projections. The assessed exposure based on the regionalised approaches is higher than that derived 24 from the reference approach by up to 60 million people (39%). Accounting for urbanisation and coastal migration 25 leads to an increase in exposure, whereas considering urban sprawl leads to lower exposure. Differences between 26 the reference and the regionalised approaches increase with higher SLR. The regionalised approaches show highest 27 exposure under SSP5 over most of the 21st century, although total population in SSP5 is the second lowest overall. 28 All methods project the largest absolute growth in exposure for Asia and relative growth for Africa.

Keywords: sea-level rise, Shared Socioeconomic Pathways, coastal population dynamics, coastal flooding
 exposure

31 1 Introduction

A large number of studies have assessed future coastal exposure to sea-level rise (SLR) and respective impacts on global scale (e.g. Hanson et al. 2011; Hallegatte et al. 2013; Neumann et al. 2015). These studies rely on SLR and socio-economic scenarios, because future climate and socio-economic change cannot be forecasted over decades due to deep uncertainties and alternating pathways of development involved. While a lot of emphasis has been placed on developing adequate SLR scenarios that account for uncertainties in future SLR, much less emphasis has been placed on socio-economic scenarios, even though both uncertainties are roughly at equal footing in terms of their influence on future coastal exposure and impacts (Hinkel et al. 2014).

39 The implementation of population changes in global coastal impact assessments has generally improved since the 40 1990s, as at that time studies assumed socioeconomic conditions to remain constant (e.g. Nicholls and Mimura 41 1998) and were therefore unrealistic for future conditions. In recent years multiple scenarios of socioeconomic 42 development on global, continental or national level have been employed in global coastal impact assessment in 43 order to account for uncertainties in socioeconomic development and lead to plausible estimates on future exposure 44 (see e.g. Nicholls (2004) and Arnell et al. (2004) for the Intergovernmental Panel on Climate Change (IPCC) Special Report on Emission Scenarios (SRES) and e.g. Hinkel et al. (2014) for the Shared Socioeconomic 45 Pathways (SSPs)). However, these approaches used population projections on national level and did not account 46 47 for different population change rates in coastal and inland areas. As coastal zones typically face different 48 challenges compared to inland areas, including differing rates of economic growth and a higher density of cities 49 (McGranahan et al. 2007; Seto 2011; Kummu et al. 2016), coastal population was underestimated.

50 For this reason, some recent studies of global coastal exposure have used higher growth rates for coastal population 51 than for inland population. Nicholls et al. (2008) assumed coastal population to grow up to 2 times faster than the 52 national average. Neumann et al. (2015) refined the approach of Nicholls et al. (2008) and differentiated between 53 coastal and inland population development for urban and non-urban areas by using correction factors. These 54 corrections factors allowed coastal population to remain constant if inland population was projected to decrease 55 and grew 1.7 to 2 times faster than the inland population if the inland population was projected to increase. These 56 approaches have the limitations of assuming first, arbitrary correction factors, and second that coastal population 57 develop faster than inland population for all countries, which is, not always the case. Merkens et al. (2016), for 58 example, tested this assumption against historical population data for coastal countries between 1990 and 2000 59 and found that for 40-50% of all countries inland urban and rural locations grow faster than their coastal 60 counterparts.

61 Spatially explicit population projections provide a more realistic basis for coastal exposure analysis. Gaffin et al. 62 (2004) developed population projections until 2100 consistent with the SRES with a horizontal resolution of 15' 63 (~30 km at the equator). Grübler et al. (2007) produced gridded population projections with a horizontal resolution 64 of 7.5' (~15 km at the equator) for three of four SRES scenarios. Their work was refined by Jones and O'Neill 65 (2016), who created gridded population projection for all five SSPs at an initial horizontal resolution of 7.5'. Their 66 projections were downscaled to 0.5' (~ 1km at the equator) by Gao (2017). Jones and O'Neill (2016) analysed 67 historical trends of population development and used a gravity-based downscaling model to simulate urban and 68 rural population changes. For all five SSPs an index of potential attractiveness for each grid cell was used to 69 allocate population, which indirectly leads to different growth rates on subnational level for coastal and inland 70 areas. Merkens et al. (2016) created gridded population projections with a horizontal resolution of 0.5' for all five 71 SSPs that focused on coastal areas and analysed historical growth differences of coastal urban and coastal rural 72 areas compared to the inland counterparts. Their method is described in more detail in section 2.2. In addition, 73 they expanded the qualitative narratives of the SSPs to the coastal zone and assumed scenario-specific 74 modifications of the observed growth differences that are based on the narratives. Both studies, Jones and O'Neill 75 (2016) and Merkens et al. (2016), are consistent with the population projections (KC and Lutz 2017) and 76 urbanisation projections (Jiang and O'Neill 2017) on national level that are used in the SSP framework (O'Neill 77 et al. 2017).

78 In this study we assess the sensitivity of outcomes in coastal exposure analysis to inclusion of subnational 79 heterogeneity in population projections. We compare (i) homogeneous population change on national level 80 (hereinafter referred to as the basic approach) to (ii) the population projections of Merkens et al. (2016) that also 81 account for urbanisation and coastal migration are have been specifically developed for coastal exposure analysis 82 (hereinafter referred to as the coastal approach) and to (iii) the downscaled spatial projections of Jones and O'Neill 83 (2016) by Gao (2017) that account for urbanisation and urban sprawl (referred to as dynamic approach)). We 84 further analyse (iv) the extent to which urbanisation can explain the differences in exposure between the basic and 85 coastal approach (referred to as urban approach) (see Fig. 1).



Fig. 1: Regionalisation approaches. The basic approach assumes homogeneous population dynamics within a country. The
coastal approach differentiates population dynamics between coastal urban, coastal rural, inland urban and inland rural
areas. The dynamic approach uses dynamic urban extents to account for urban sprawl. The urban approach differentiates
urban and rural population dynamics with static urban extents

91 2 Data and Methods

92 2.1 DIVA database

For our analysis, we employ the Dynamic Interactive Vulnerability Assessment (DIVA) modelling framework,
which has been used in a wide range of applications in coastal risk assessments (see Hinkel et al. 2013 for erosion,
Hinkel et al. 2010 and Hinkel et al. 2014 for adaptation, Hinkel et al. 2012 for adaptation and mitigation, Spencer
et al. 2016 for wetlands). The results presented in this study are based on version 30 of the DIVA database and
model version 1.7.

- 98 The DIVA database breaks the world's coasts (excluding Antarctica) into 12,148 segments. Each coastal segment
- 99 provides information on administrative, bio-physical and socioecological attributes. In the context of this study,
- 100 we focus on the population living in the 1 in 100-year floodplain, which is a well-established measure of coastal
- 101 exposure analysis (e.g. Hanson et al. 2011; Vousdoukas et al. 2016; Muis et al. 2017). The 1 in 100-year coastal
- 102 flood heights are taken from Muis et al. (2016). We use the DIVA flood module to calculate the number of people

living in the floodplain without considering dikes. A detailed description of this approach can be found in Hinkel
et al. (2014). As we account for isostatic adjustment and subsidence (see section 2.3), DIVA provides relative sealevel for all segments.

106 To define the floodplain, we use a global elevation dataset which is based on SRTM (Jarvis et al. 2008) and 107 GTOPO30 (USGS 1996) data for high latitudes (> 60° N and > 54° S). For all elevation steps from 1m to 16m, we 108 calculate the extent of the area that is hydrologically connected to the ocean and smaller or equal to the respective 109 elevation threshold (see Poulter and Halpin 2008). Intermediate values are linearly interpolated (Hinkel et al. 2014). We utilise the GRUMPv1 grid (CIESIN et al. 2011a) to analyse the population located in each of these 110 111 elevation increments for the year 2000. The coastal SSPs of Merkens et al. (2016) use the GRUMP urban extent 112 grid, which uses census population counts, settlement points and night-time lights, to define urban areas (CIESIN 113 et al. 2011b) and assumes assume these to be static. GRUMP tends to underestimate the extent of settlements with 114 none or little light at night, e.g. in parts of Africa (Balk et al. 2006), which also affects the estimates on exposed 115 population. The estimates on exposed population also depend on the elevation model used for the analysis. Lichter 116 et al. (2011) analysed the land area of the LECZ derived from three different elevation datasets with the same 117 vertical and horizontal resolution of 1m and 0.5' (~ 1km at the equator). On continental scale, they found differences of up to 40%. In the same study, Lichter et al. (2011) compared two commonly used population datasets 118 119 (GRUMP alpha and LandScan 2006) and analysed the population located in the LECZ. On global scale, the LECZ 120 population differed by $\sim 10\%$, on continental scale by up to 28%. They stated the combined uncertainty of elevation 121 and population data at 20% on global scale and at up to 67% on continental scale. Mondal and Tatem (2012) compared the LECZ population for GRUMP version 1 (the same version that was used in this study) and LandScan 122 123 2008 and found differences of 4% on global scale and of up to 39% on continental scale. GRUMP's underlying 124 assumption of homogeneous population distribution within urban and rural areas in the same administrative unit 125 can in addition lead to an over- or underestimation of the 'true' exposure (Merkens and Vafeidis 2018). As this 126 study uses the same population and elevation datasets throughout the analysis, we expect the relative differences 127 between the approaches to be independent from the elevation or population data, whereas the absolute numbers 128 are likely to be different if other population or elevation data are used.

129 2.2 Socioeconomic scenarios

We initially calculate exposure of population based on two approaches to account for future population development in coastal areas (see Fig. 1). In the basic approach, we use national population projections taken from KC and Lutz (2017) and apply these to the baseline (i.e., year 2000) spatial population data. This approach assumes 133 homogeneous growth rates within each country, i.e. population in coastal areas grows at the same rate than in 134 inland areas. In the coastal approach, we use the coastal SSPs of Merkens et al. (2016). These are based on the 135 national population projections of KC and Lutz (2017) as well, but consider urbanisation projections (Jiang and 136 O'Neill 2017), historical growth differences and scenario-dependent modifications of growth differences. For each 137 country Merkens et al (2016) analysed the population growth for coastal urban (rural) areas and inland urban 138 (rural) areas over a 10-year period from 1990 to 2000. If coastal areas had a higher population growth rate than 139 inland areas, the growth difference (GD) was positive and vice versa. The GD allows for negative (positive) 140 population growth in the coast or inland even if national population growth is positive (negative). It also allows 141 for higher population change rates in coastal areas compared to inland areas. For SSP2 Merkens et al. (2016) 142 assumed the GD to keep constant over time for each location. For the other four SSPs they modified the GDs based 143 on the interpretation of the coastal SSP narratives, which are introduced in the same study. They quantified the 144 modification of the GDs based on the difference between percentiles in the distribution of the observed urban and 145 rural GDs for all coastal countries. In SSP1 they assume no differences in growth for coastal and inland urban 146 areas and a reduced rural GD (translates to relatively higher rural growth in inland). In SSP3 they assume that the 147 GD to reduce by 50% for both, urban and rural areas. In SSP4 and SSP5 they increased the GD (translates to 148 relatively higher relative growth at the coast), whereby the increase was bigger in SSP5. Based on the scenario 149 specific GDs and the population and urbanisation projections they calculated population counts for coastal urban, 150 costal rural, inland urban and inland rural for each country in 5 year increments until 2100. This leads to 151 heterogeneous growth rates within countries because urban areas develop differently to rural areas and coastal 152 areas differently to inland areas. We then calculate the mean coastal population growth rate for each country and 153 apply it on each coastline segment of this country. We must note that the definitions of 'urban' between GRUMP 154 (used in Merkens et al. (2016) and Gao (2017)) and Jiang and O'Neill (2016) differ, which results in an offset in 155 the data for the years 2005 and 2010 (see section 4 for a discussion of the implications on exposure analysis).

156 2.3 Sea-level rise scenarios

We use the projected changes in global mean sea level and the likely ranges reported in the Fifth Assessment Report of the Intergovernmental Panel on Climate Change (Church et al. 2013). For each of the four RCPs, we use the ensemble median as medium SLR scenario. The 83rd percentile serves as high SLR scenario and the 17th percentile as low SLR scenario (see Table 1). We do not consider regional patterns of SLR due to ocean dynamics and regionally differential changes in thermal expansion and rotational and gravitational effects of the mass loss of ice sheet. Church et al. (2013) show that these regional effects are below 10% for most of the populated coastal zone with the exception of the East Coast of the US. Hence the global effects of these regional SLR variations are expected to be much smaller than those of human-induced subsidence in densely populated river deltas, which we consider here together with isostatic adjustment. Furthermore, uncertainties in regional sea level projections are large, with different models producing different patterns and the highest deviations of regional sea-level rise due to dynamic variability coinciding with those regions for which model uncertainties are largest (Church et al., 2013). We assume that water levels during coastal floods increase by the same amount as the projected global sea-level and do not account for non-linear interactions between the water level and SLR (Arns et al. 2017) as the focus of this paper is the comparison of population distribution approaches.

171 Table 1: Sea level rise projections for 2100 referenced to the 1986-2005 period [in m].

	low	medium	high
RCP2.6	0.28	0.44	0.61
RCP4.5	0.36	0.53	0.71
RCP6.0	0.38	0.55	0.73
RCP8.5	0.53	0.74	0.98

Values are taken from Prather et al. (2014).

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In this study, we use the 12 SLR scenarios from Table 1 (four RCPs, for each high, medium and low SLR 173 174 projections). These are combined with the five SSPs. Taking into account the two regionalisation approaches (plus 175 another two for testing our assumption) in each SSP, we end up with 240 model runs. This number could be reduced by ignoring scenario combinations that are not plausible. For example, the combination of an 176 177 environmentally friendly socioeconomic scenario (SSP1) and a physical scenario with high radiative forcing (RCP8.5) would in general be inconsistent (van Vuuren et al. 2014; Engström et al. 2016). Nevertheless, we 178 179 decided to analyse all scenario combinations, as this study aims to analyse and understand the effect that 180 regionalisation approaches of socioeconomic scenarios have for impact assessment.

181 3 Results

We compare future coastal exposure to 1 in 100-year coastal floods based on the different regionalisation approaches. We define the absolute difference in exposure as the difference in the tested approach (i.e. coastal, urban or dynamic) minus the exposure in the basic approach. The relative difference is defined as the absolute difference in exposure divided by the exposure in the basic approach.

186 3.1 Global

Our first main finding is that accounting for urbanisation and coastal migration has significant implications for assessing coastal exposure. The exposure based on the coastal approach exceeds the one based on the basic approach in all scenarios over the 21st century (see Fig. 2). This finding is consistent for all SLR scenarios (see

190 Fig. A. 3). For SSP1, 4 and 5 we find the exposure in the basic approach with high SLR in all RCPs to be lower 191 than the respective low SLR variant in the coastal approach. In other words, in these scenarios the difference 192 between the population distribution approaches is larger than the difference between high and low SLR. To 193 investigate which of the two (urbanisation and coastal migration) is the dominant process leading to the difference 194 between basic and coastal approach, we added the 'urban approach' to our modelling scheme (see Fig. 1). The 195 urban approach is based on population and urbanisation projections that are modelled in the same way as in the 196 coastal SSPs, but uses a GD of zero, which means that the population in urban and rural zones for each SSP grows 197 at rates consistent with projections on national level and does not differ between coastal and inland areas. We 198 assume that the difference between the urban approach and the basic approach represents the impact of changing 199 urbanisation levels, without considering urban sprawl. The difference between the urban approach and the coastal 200 approach can result from differences in fertility, mortality, international migration or internal migration, of which 201 we assume internal migration from or to the coast to have the highest impact. We find that, independently of SLR, 202 urbanisation explains 61% of the difference between the coastal and basic approach in SSP1, 96% in SSP2, 54% 203 in SSP3, 76% in SSP4 and 45% in SSP5 (see Fig A. 2). This means that SSP5 is the only scenario where 204 urbanisation appears not to be the dominant process. This can be explained by the underlying assumptions of 205 intense coastward migration for SSP5 in the coastal approach (Merkens et al., 2016). In general, the projected 206 increase in urbanisation levels leads to higher population growth rates in the coastal zone compared to inland areas, 207 as coastal areas show a higher density of cities than inland areas, and population is projected to move into these 208 cities. In the basic approach, the population in all areas within a country grows at the same rate, which leads to 209 lower population numbers at the coast compared to the coastal approach. We therefore conclude that the higher 210 exposure in the coastal approach compared to the basic approach is due to a combination of increasing urbanisation 211 levels in all SSPs and migration to coastal areas, of which urbanisation is the dominant process for SSPs 1-4 and 212 coastal migration for SSP5.



213

214 Fig. 2: Exposure of population to 1 in 100-year coastal floods under medium SLR in RCP6.0 in the tested approaches.

215 Our second main finding is that the implementation of urban sprawl has a considerable impact on the estimates on 216 exposure. We compare the urban approach to a 'dynamic approach', which is based on the population projections 217 of Jones and O'Neill (2016) that were downscaled by Gao (2017). Unlike the urban (and coastal) approach, that 218 assume urban extent to be static, the dynamic approach considers urban sprawl, which leads to wider city extents 219 and lower population densities within cities. We assume that differences between the urban and dynamic approach 220 are mainly due to urban sprawl, as the approaches use the same population projections of KC and Lutz (2017) and 221 the same urbanisation projections of Jiang and O'Neill (2017). Compared to the dynamic approach, we find 222 exposure to 1 in 100-year coastal floods to be higher in the urban approach for all combinations of SSPs and RCPs

223 (Fig. A. 3). They differ between 15 million in SSP1 (RCP 2.6 and low SLR) and 26 million in SSP4 (RCP 8.5 and 224 high SLR). Differences in SSP1 are lowest, as cities in the dynamic approach are assumed to be concentrated 225 (Jones and O'Neill 2016) and urban extents to be static in the urban (and coastal) approach. However, the 226 difference of 15 million in SSP1 is considerable and suggests that the definition of urban areas (and population) 227 between the urban and the dynamic approach differs, as urbanisation levels and total population do not differ and 228 cities are assumed to be concentrated (dynamic approach) or static (urban approach). For SSPs 2-5 differences 229 between the urban approach and dynamic approach are higher, as only the dynamic approach considers urban 230 sprawl. This suggests that urban sprawl can lead to a reduction of exposure as cities seem to expand towards less 231 flood-prone areas. The differences between the basic and the dynamic approach are rather small (Fig. 2). Global 232 exposure in the dynamic approach under SSP3 for 2100 is up to 7.5 million lower than one in the basic approach. 233 In the other SSPs, exposure based on the dynamic approach exceeds the basic approach by 1 million in SSP4, 2 234 million in SSP2, 5 million in SSP5 and 6 million in SSP1 (see Fig. A. 4). These SSPs are also projected to have a 235 high increase in urbanisation levels, whereas urbanisation levels in SSP3 are projected to increase little (Jiang and 236 O'Neill 2017). This supports our first finding that neglecting urbanisation patterns would lead to an 237 underestimation of coastal exposure. The differences between the dynamic and the coastal approach are larger than 238 the differences between the dynamic and the urban approach (between 17 million is SSP2 under RCP 2.6 with low 239 SLR and 54 million in SSP5 under RCP 8.5 and high SLR), as coastal migration is additionally considered in the 240 coastal approach. Overall we believe that the coastal approach overestimates exposure, as it does not consider 241 urban sprawl, which appears to reduce exposure; and that the dynamic approach underestimates exposure, as it 242 does not explicitly consider coastal migration, which appears to increase exposure to coastal flooding. We must 243 note that this study does not aim to test the underlying quantifications on coastal migration in Merkens et al. (2016) 244 and the quantification of urban sprawl in Jones and O'Neill (2016), but rather to investigate the implications for 245 coastal exposure analysis when accounting or neglecting of processes actually taking place in coastal areas.

246 We also find that the population distribution approach is important in determining which SSP leads to the highest 247 exposure to coastal flooding. Though all approaches agree on SSP3 having the highest exposure in 2100, only the 248 basic approach shows SSP3 to lead to the highest exposure throughout the century. The other approaches agree on SSP5 leading to the highest exposure until 2060 (dynamic approach), 2075 (urban approach) and 2090 (coastal 249 250 approach) (see Fig. 2). This holds true for all SLR scenarios. This is noteworthy as SSP5 and SSP1 are projected 251 to have considerably lower total populations than the other SSPs (KC and Lutz 2017). We identify two factors 252 leading to this observation. The behaviour in the basic approach can be explained by the underlying global 253 population projections that project population to be highest in SSP3 (KC and Lutz 2017). The higher exposure in SSP5 in the other approaches is due to high urbanisation levels (Jiang and O'Neill 2017). Exposure rises in the coastal approach as coastal areas are assumed to be more attractive than inland areas and decreases in the dynamic approach as high urban sprawl leads to cities expanding to flood proof areas.

257 Results also show that the absolute difference in exposed population between the basic and the other approaches 258 increases with SLR (see Fig. A. 4). We find the highest differences under the high SLR projections in RCP8.5 and 259 the smallest differences under the low SLR projections in RCP2.6. Compared to the basic approach, SSP1, SSP4 260 and SSP5 show the highest difference and SSP2 and SSP3 the lowest. Different to the urban and the coastal approach, the dynamic approach shows a reduced exposure for SSP3 and a higher difference for SSP2 than for 261 262 SSP4 for 2090 to 2100, when the basic approach is used as reference. Again, this observation highlights the 263 significance of urbanisation, coastal migration and urban sprawl.. As cities are concentrated in coastal areas, the 264 overall population growth in coastal areas is higher than the national average (represented by the basic approach).

265 3.2 Regional

In this section we focus on the comparison between the basic and the coastal approach, as the coastal approach explicitly considers coastal migration. Results for the dynamic and urban approach on regional level can be found in the SM.

269 Different to the global patterns, Europe, Northern America and Oceania face the highest exposure under SSP5 for 270 both coastal and basic SSPs (Fig. 3). Exposure increases continuously until 2100 under this SSP. For Africa and 271 Latin America and the Caribbean (LAatC), SSP3 shows the highest exposure throughout the century, which also 272 increases continuously with time. This is in line with the underlying national projections of KC and Lutz (2017) 273 that project highest population under SSP5 for the most developed countries and under SSP3 for developing 274 countries. For Asia, we find a notable difference between the coastal and basic approach. In the basic approach, 275 exposure is highest under SSP3 throughout the 21st century. In the coastal approach, exposure is highest under 276 SSP5 until 2075 and under SSP3 afterwards. Asia's high exposure under SSP5 in the coastal approach reflects the high increase of urbanisation levels in the underlying urbanisation projections (Jiang and O'Neill 2017) and the 277 278 coastward migration in the coastal SSPs. The decrease in Asia's exposure projected after 2050 is due to the 279 decreasing population after 2050 in the underlying population projections (KC and Lutz 2017). This also can also be seen in the basic approach and holds true for all SSPs except SSP3, where the Asia's population is projected to 280 281 grow after 2050.



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Fig. 3: Exposure per continent under medium SLR in RCP6.0. In the basic approach, SSP1 and SSP5 overlap for Africa and
Asia. For LAtaC (Latin America and the Caribbean), SSP1 and SSP4 overlap in both approaches.

285 The absolute difference in exposure to 1 in 100-year coastal floods on continental scale follows global patterns

and becomes larger with SLR in all SSPs (see Fig. A. 7). Accordingly, we find the highest differences in RCP8.5

with high SLR and the smallest differences in RCP2.6 with low SLR. The difference between the coastal and basic
approach is highest in SSP5 in all continents except Africa, where SSP4 shows the highest differences. We observe
the lowest differences in SSP1 for Africa, LAatC, Northern America and Oceania. For Europe and Asia, we find
the lowest differences between the coastal and basic approach in SSP3.

291 The relative difference in exposed population is heterogeneous and does not follow the global patterns. For Africa, 292 which shows overall the highest values, we find the relative difference to decrease with rising sea levels (see Table 293 A. 1). The highest difference in exposure is in SSP4 (coastal approach is up to 64% higher than the basic approach) and lowest in SSP2 and SSP3 (coastal approach 24% higher than basic approach). For Asia, we find the highest 294 295 relative differences between coastal and basic approach in SSP5 (48%) and the lowest in SSP3 (~12%). For 296 Europe, which shows overall the closest agreement between coastal and basic approach, the relative difference in 297 exposure increases slightly with SLR. SSP5 exhibits the highest relative difference in exposure (~20%) and SSP3 298 the lowest (<1%). For Northern America, the relative difference in exposure increases with SLR in SSP1 and 299 decreases in SSP2-5 while the opposite is the case in Asia. For LAatC and Oceania we do not find a relation 300 between SLR and relative difference in exposure based on the basic and coastal SSPs.

301 4 Discussion

302 One of our key findings is that under all scenarios the coastal approach projects higher population located in the 303 floodplain of 1 in 100-year coastal floods than the basic approach. In agreement with previous studies that 304 identified urbanisation as a key component in coastal population development, we explain most of the differences 305 with the projected growing urbanisation levels in the coastal approach (see Fig. A. 1 and Fig. A. 2). Coastal areas 306 today show a higher concentration of cities than inland areas. Kummu et al. (2016) shows that 105 out of 256 cities with a population of more than 1 million are located in the near coast zone (proximity to coast < 100 km and 307 308 altitude < 100 m). According to Brown et al. (2013), in 2010 20 out of 31 megacities (cities with more than 8 309 million inhabitants) were located in the low-elevation coastal zone (LECZ; altitude ≤ 10 m and hydrological 310 connection to the ocean). Neumann et al. (2015) assume that the number of megacities in the LECZ will increase to 25 until 2025. Hoornweg and Pope (2016) project the population development of the 101 largest cities under 311 312 three SSPs. They show that the percentage of population living in these cities will increase from 11% in 2010 to 313 15% in SSP3, 20% in SSP2 and 23% in SSP1 until 2100. As the coastal approach accounts for urbanisation 314 (Merkens et al. 2016) and the basic approach does not, coastal population tends to be underestimated in the basic 315 approach.

316 The basic approach shows similar results to the study of Jongman et al. (2012) that also used a homogeneous 317 population growth approach on national level. They found an increase in population exposure to 1 in 100-year 318 coastal floods between 2010 and 2050 of 25% on global scale. In the basic approach, we find an increase of 319 population's exposure to 1 in 100-year coastal floods between 19% in SSP4 and 28% in SSP3. The exposure based 320 on the coastal approach grows from 2010 to 2050 between 33% in SSP3 and 50% in SSP5 and exceeds the 321 projections of Jongman et al. (2012). In agreement with Jongman et al. (2012), both approaches analysed in context 322 of this paper project the highest absolute growth in exposed population until 2050 for Asia and the highest relative 323 growth for Africa. However, the comparison of results to other studies proofs difficult, as the underlying population 324 projections are different. For example, Jongman et al. (2012) used the medium Fertility projection of the 2006 325 Revision by the UN Population Division while this study is based on the work of KC and Lutz (2017).

326 The differences in population exposure between the approaches for the years 2005 and 2010 are due to using 327 differing definitions of 'urban' in the underlying data. The urbanisation projections rely on Jiang and O'Neill 328 (2017), which used the world urbanisation prospects (UN 2015) as input data that retains the urban definitions 329 used by each country. Across countries, the definitions are inconsistent. The coastal SSPs of Merkens et al. (2016) 330 used the GRUMP urban extents grid, which tends to underestimate urban extents in developing regions (see section 2.1). Hence, urban population is concentrated in the remaining settlements with night-lights, leading to higher 331 332 estimated population counts in these areas. As coastal areas in eastern and northern Africa are heavily populated 333 (Hinkel et al. 2012) and western Africa hosts important port cities with growing population (Hanson et al. 2011), 334 the inconsistencies in data trigger an offset in the initial exposure. In SSP4, which shows the highest relative 335 differences between the coastal and basic approach for Africa, the African population grows more than threefold 336 (KC and Lutz 2017) and the urbanisation level almost doubles until 2100 (Jiang and O'Neill 2017). This leads 337 presumably to an overestimation of exposure in the coastal approach. With SLR, the effects of the initial 338 inconsistencies in the data decrease, leading to a reduction of the relative differences of exposed population.

This study has focused on the differences in exposure that arise from using different approaches to regionalise population projections. We interpret the differences in exposure between the approaches as uncertainty that is related to regionalisation, as the underlying population projections on national level do not differ between the approaches. Other uncertainties arise from elevation data and the base year population datasets used to assess the exposure to 1 in 100-year. Elevation and population datasets can potentially be improved if data availability improves and the need for modelling decreases. The uncertainties that arise from the downscaling approach can be reduced to some extent, if the differences between reported urbanisation level and the urbanisation levels based 346 on remote sensing products find a better agreement. Other parts of the uncertainty cannot be removed, as the 347 projections are made for long timeframes and human behaviour cannot be predicted.

348 5 Conclusion

349 This study compared different approaches to account for population change in coastal impact assessment in order 350 to assess the exposure of population to 1 in 100-year coastal floods under different SLR and socioeconomic 351 scenarios. All approaches were based on the same population projections on national level. We found that urbanisation and coastal migration lead to increased exposure whereas urban sprawl leads to reduced exposure. 352 353 This emphasises the need for taking into account population dynamics on subnational level in exposure assessments. We believe that the exposure estimates obtained from approaches accounting for regional variations 354 355 in population distribution, such as urbanisation, coastal migration and urban sprawl, are more reliable than the approaches not accounting for such variations. As coastal areas host a disproportionately large number of cities, 356 357 sub-national population dynamics are of particular relevance for coastal exposure studies and should not be 358 ignored. With rapidly growing cities in developing countries, the need to provide improved assessments of 359 population exposure to coastal flooding is important for global and national planning, both in terms of allocating 360 human and financial resources on national level and climate change adaptation funding on international level.

Acknowledgments: The authors would like to express heir thanks to the editor and the reviewers for their valuable
 comments. We also would like to thank Maureen Tsakiris for illustrating Fig. 1.

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