

I don't know, are you sure you want to do this?

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² **Abstract.** ...

1. Introduction

2. Sea level projections (PG)

2.1. Global sea level

3 Most climate models do not explicitly provide sea level as an output of the calculations.
4 Rather, the IPCC AR5 report [*Stocker et al.*, 2013, ch. 13] combines the heat expansion
5 of the ocean with temperature forced models for glacial melt, Greenland ice melt, and
6 Antarctic ice melt and with land rise due to rebound from the last ice age. Judging from
7 the supplementary material to *Stocker et al.* [2013, ch. 13], the uncertainty assessment is
8 only based on the spread of the ensemble of temperature projections, not on the additional
9 uncertainty in the ice models used.

10 We will instead use the empirical approach of Rahmstorf and collaborators [*Rahmstorf*,
11 2007; *Rahmstorf et al.*, 2011], employing the statistical modeling of *Bolin et al.* [2014] to
12 relate global annual mean temperature anomalies [*Hansen et al.*, 2001] to global mean sea
13 level anomalies [*Church and White*, 2011].

14 We then apply the estimated historical relationship to projected temperatures from the
15 CMIP5 experiment [*Taylor et al.*, 2012] to obtain projected global annual mean sea level,
16 taking into account the uncertainty in the statistical model as well as the spread of the
17 temperature projection ensemble (see subsection 2.3).

2.2. Local sea level

18 In order to get from global sea level projections to local ones, it is important to note
19 that sea level rise not is uniform over the globe. Glacial and land ice melting affect the

local sea level differently depending on where the melted ice is located. Another major effect in Fennoscandia is the land rise due to isostatic rebound from the glaciers of the last ice age. Again, we will use historical data to relate global sea level to isostatically corrected local sea level using a time series regression model. Figure 1 shows uncorrected and corrected Bergen sea level data, and the relationship between the corrected Bergen data and the global sea level data. The time series regression uses an ARMA(1,1)-model [Box and Jenkins, 1970].

The local sea level projections are then obtained by first relating projected temperature to global sea level, and then relating the global sea level to the local one. Each climate model temperature projection yields a different local sea level projection.

2.3. Uncertainty assessment

Following the approach of Guttorp *et al.* [2014] we assess the uncertainty in the local sea level projections taking into account the variability between the climate projections used, the uncertainties in the regressions of global mean temperature on global mean sea level and of global on local sea level. We express the sea level projection uncertainty in terms of a confidence band that is simultaneously of the intended level for all projection years. This allows us, for example, to get a confidence band for the years when a given sea level rise is obtained.

2.4. Limitations of the sea level projections

The main assumption is using historical relationships in statistical projections of the type used in this paper is that there is no major change in how temperature relates to sea level, globally and locally. Among the factors that may invalidate this approach are

changes in water storage on land (removing water from the oceans), excessive siphoning of groundwater (resulting in land subsidence), changes in the rates of glacial and land ice melt, and changes in Earth's gravitational field due to transfer of mass from land ice to ocean water. For example, the rate of ice melt on Greenland may suddenly increase substantially due to intense warming of both air and sea water. Our current climate models are not able to resolve the ice processes sufficiently to include such so called tipping points into the projections. Also, the IPCC scenarios [*van Vuuren et al.*, 2011] do not include changes in water usage.

3. Decision tools (KdB, MD, TT)

3.1. Timing of adaptation measures

We consider adaptation decision making related to the timing of proactive adaptation measures. That is, the goal is to adapt to sea level rise before major damages occur. In a cost-benefit framework, an investment should be delayed as long as the benefits of delay (avoided investment costs) are greater than the associated costs (higher climate change damages) [*Fankhauser et al.*, 1999].

Fankhauser et al. [1999] describe a deterministic framework where an adaptation investment of C^0 now (at time $n = 0$) leads to unmitigated damage of d_0^0 in period 0, and a stream of partially mitigated damages d_t^0 in periods $t = 1, 2, \dots$. If r is the discount rate, the net present value damage, D^0 , associated with this investment is

$$D^0 = C^0 + d_0^0 + \frac{d_1^0}{1+r} + \frac{d_2^0}{(1+r)^2} + \dots \quad (1)$$

In comparison, postponing the adaptation investment to time period $n = 1$ would lead to unmitigated damages in periods 0 and 1, and partially mitigated damages, d_t^1 , thereafter.

The delay would be preferable if

$$C^0 - \frac{C^1}{(1+r)} > (d_0^1 - d_0^0) + \frac{d_1^1 - d_1^0}{1+r} + \frac{d_2^1 - d_2^0}{(1+r)^2} + \dots$$

Here, the expression on the left describes the benefits of the delay while the expression on the right describes the cost of the delay. In the simplest case, there is no change in investment costs ($C^0 = C^1 = C$) and the delay has no lasting effects beyond period 1 ($d_t^1 = d_t^0$ for $t > 1$). In this case, the comparison is between the expected return r earned on the capital while implementation is delayed and one additional time period of unmitigated damage,

$$rC > d_1^1 - d_1^0.$$

3.2. Limitations of the decision framework

4. Case studies

4.1. Data (PG)

We use local tide gauge data from the Permanent Service for Mean Sea Level, UK, which is the worldwide repository for national sea level data. Climate projections are from the fifth climate model intercomparison project, CMIP5.

4.2. Timing of adaptation measures (KdB, TT)

A case study focusing on and comparing different cities in Norway.

4.3. Selection of adaptation measures(?) (MD)

A case study focusing on Denmark.

5. Conclusions

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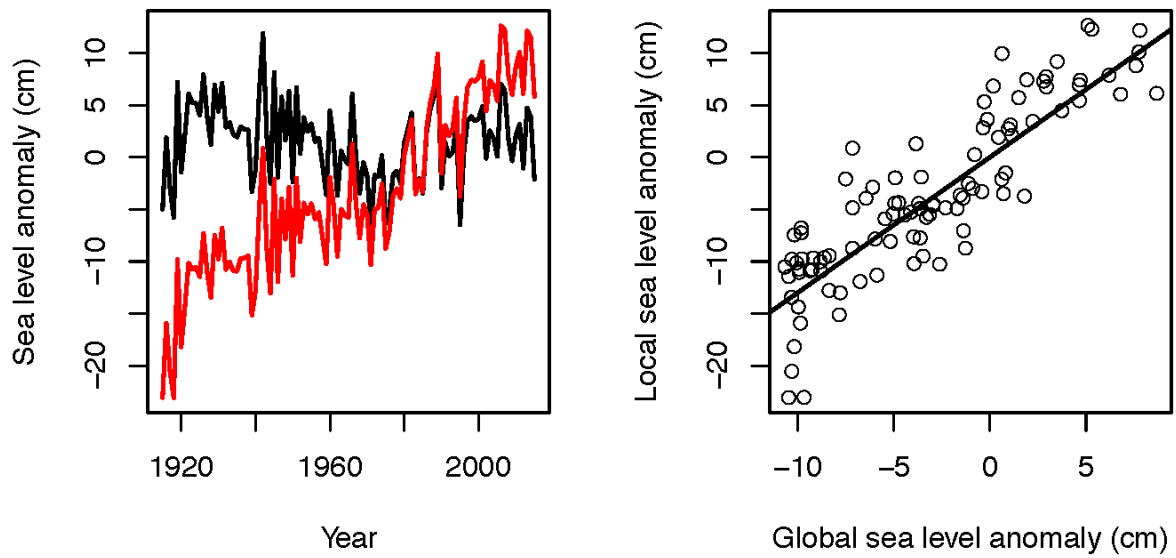


Figure 1. The left figure shows raw (black) and gia-corrected (red) sea level data from Bergen, The right figure relates the gia-corrected Bergen sea level to the global sea level series of *Church and White* [2011]. The straight line is the time series regression line.

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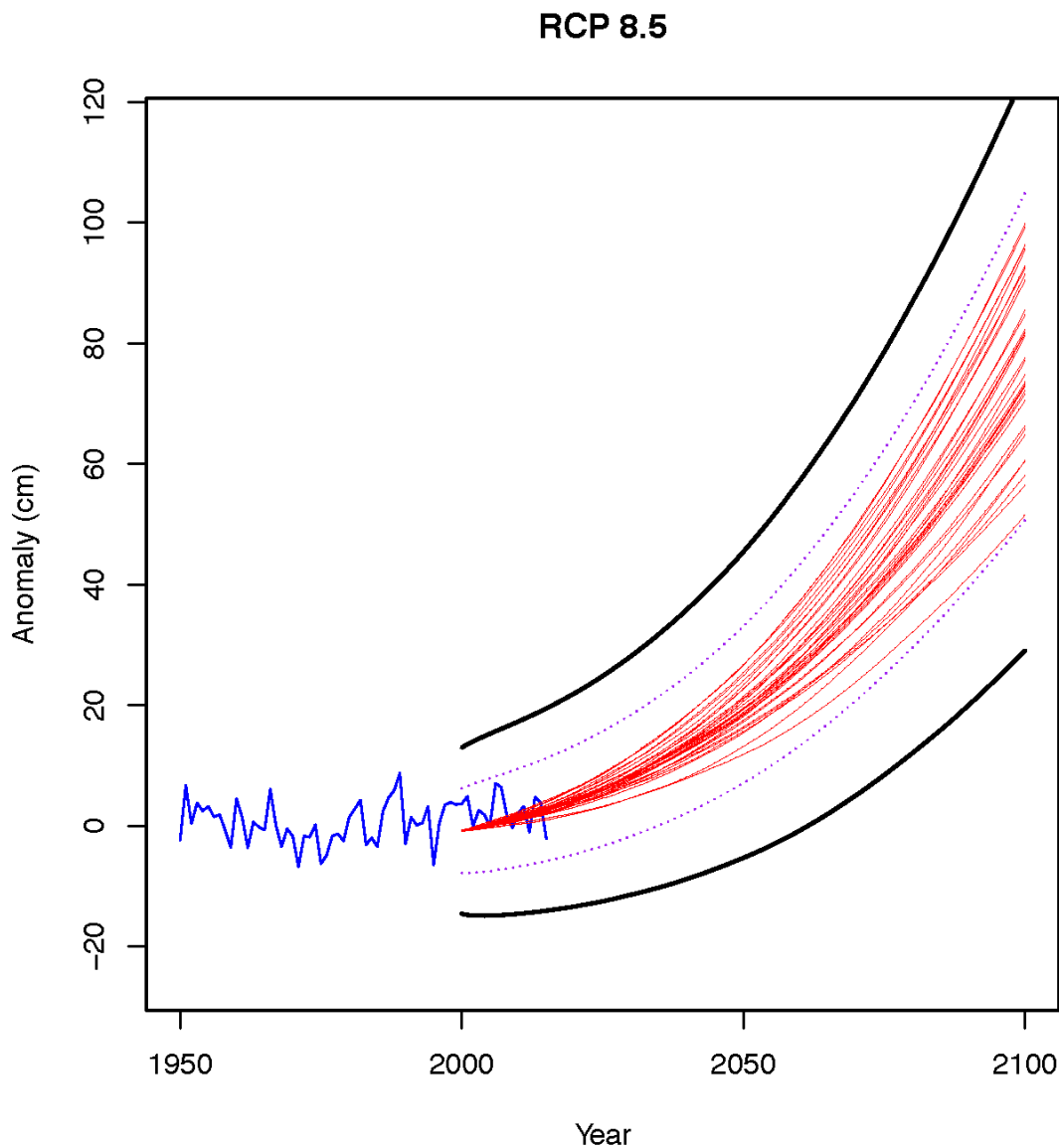


Figure 2. Simultaneous 90% confidence set (thick black lines) for Bergen sea level projections for the years 2000-2100 using RCP8.5. The sea level data are shown in blue and end in 2015. The thin red lines are the projections without uncertainty based on each of the climate models. The dashed purple lines connect pointwise confidence intervals for each year.