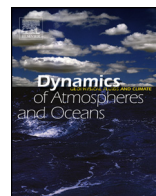




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## The statistical relation of sea-level and temperature revisited



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### ABSTRACT

We propose a semi-empirical model for the relation between global mean surface temperature and global sea-levels. In contradistinction to earlier approaches to this problem, the model allows for valid statistical inference and joint estimation of trend components and interaction term of temperature and sea-level. Estimation of the model on the data set used in [Rahmstorf \(2007\)](#) yields a proportionality coefficient of 4.6 mm/year per °C at a one-sided significance level of 7.6 percent or higher. Long-term simulations of the model result in a two-sided 90-percent confidence interval for the sea-level rise in the year 2100 of [15 cm, 150 cm] above the 1990 level. This is a wider margin of error than was reported in the previous literature, and it reflects the substantial uncertainty in relating two trending time series.

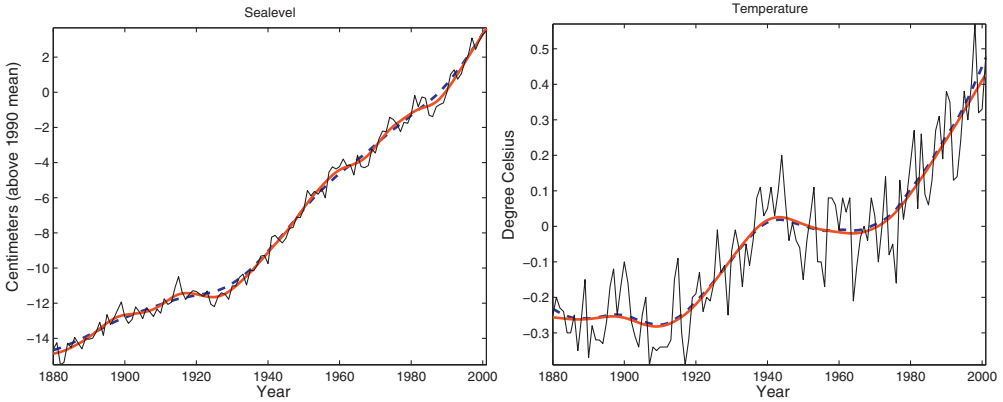
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## 1. Introduction

Estimating the influence of global temperature on global sea-level using time series of historical data is a statistical problem of observational data that has been discussed intensively in recent years

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**Fig. 1.** Trend components in global sea-level (left) and global surface temperature (right). The solid smooth line is the trend component estimated from the state-space model. The dashed smooth line is the non-linear trend component estimated from singular spectrum analysis.

(Rahmstorf, 2007a,b, 2010; Woodworth et al., 2009; Horton et al., 2008, Jevrejeva et al., 2009, 2010). A central problem is the fact that both temperature and sea-level time series are trending upwards (see Fig. 1) and that the hypothesized relationship between the two lies exactly in these long-term trends. In other words, it is supposed that the long-term trend in global temperatures influences the long-term trend in global sea-levels (Rahmstorf, 2007; Rahmstorf and Vermeer, 2009). Therefore, standard statistical methods of differencing or de-trending and subsequent regression of stationary residuals onto each other are beside the point.

Semi-empirical models have been proposed that justify a linear approximation of the interaction term of these trend components in temperature and sea-level, which make the problem's estimation accessible by linear regression methods (Jevrejeva et al., 2009; Grinsted et al., 2010; Rahmstorf and Vermeer, 2009). In Rahmstorf (2007), Rahmstorf and Vermeer (2009), a linear approximate relationship for the influence of global temperatures on the differences of global sea-levels is motivated:

$$\frac{dH}{dt} = a(T - T_0), \quad (1)$$

where  $H$  is the sea-level,  $T$  is temperature, and  $a$  is the coefficient of influence. In order to estimate this relation, a linear regression of the type

$$f(H_t) - f(H_{t-1}) = b(f(T_t) - f(T_0)) + \text{error}_t \quad (2)$$

is run, where  $H_t$  is a time series of global sea-level observations,  $T_t$  is a time series of global temperature observations,  $b$  is the coefficient of influence in the model and an estimate of  $a$  in Eq. (1). For a given time series  $X_t$ , the data transformation  $f(X_t)$  extracts a long-term trend. The extant literature often uses singular spectrum analysis in order to smooth short-run fluctuations, see for example Moore et al. (2005), Jevrejeva et al. (2006), Wahl et al. (2010), Rahmstorf et al. (2012a,b), Bittermann et al. (2013). Sometimes, moving averages over non-overlapping windows are applied in addition to singular spectrum analysis to further smooth the series (Rahmstorf, 2007; Rahmstorf and Vermeer, 2009). In the case of global sea-level and temperature time series  $H_t$  and  $T_t$  considered in Rahmstorf (2007), the regressand  $f(H_t) - f(H_{t-1})$  and regressor  $f(T_t) - f(T_0)$  are both upward trending.

The linear regression approach with trending time series potentially invalidates standard statistical inference (Schmith et al., 2007, 2012; Granger and Newbold, 1974). This is caused by the properties of the residual process ( $\text{error}_t$ ), which may be statistically indistinguishable from a random walk. For the data set considered in this paper, we show that this is indeed the case (see Fig. 4 and Table 2). Smoothing by singular spectrum analysis adds to the complexity of the statistical properties of the process ( $\text{error}_t$ ), leaving the asymptotic properties of the estimator of  $b$  in Eq. (2) intractable.

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