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Abstract

Understanding the drivers of ecosystem change and their effects on ecosystem services are essential for management decisions and verification of progress towards national and international sustainability policies (e.g. Aichi Biodiversity Targets, Sustainable Development Goals). We aim to disentangle spatially the effect of climatological and non-climatological drivers on ecosystem service supply and trends. Therefore, we explored time series of three ecosystem services in Switzerland between 2004 and 2014: carbon dioxide regulation, soil erosion prevention, and air quality regulation. We applied additive models to describe the spatial variation attributed to climatological (i.e. temperature, precipitation and relative sunshine duration) and non-climatological drivers (i.e. random effects representing other spatially structured processes) that may affect ecosystem service change.

Obtained results indicated strong influences of climatological drivers on ecosystem service trends in Switzerland. We identified equal contributions of all three climatological drivers on trends of carbon dioxide regulation and soil erosion prevention, while air quality regulation was more strongly influenced by temperature. Additionally, our results showed that climatological and non-climatological drivers affected ecosystem services both negatively and positively, depending on the regions (in particular lower and higher altitudinal areas), drivers, and services assessed.

Our findings highlight stronger effects of climatological compared to non-climatological drivers on ecosystem service change in Switzerland. Furthermore, drivers of ecosystem change display a spatial heterogeneity in their influence on ecosystem service trends. We propose an approach building on an additive model to disentangle the effect of climatological and non-climatological drivers on ecosystem service trends. Such analyses should be extended in the future to ecosystem service flow and demand to complete ecosystem service assessments and to demonstrate and communicate more clearly the benefits of ecosystem services for human well-being.

Keywords: Regulating services, remote sensing, time series, trends, climate change, land use change

1. Introduction

Ecosystem services (ES) have been defined as benefit humans obtain from ecosystems (Millennium Ecosystem Assessment, 2005) and as contribution from ecosystems to these benefits (European Commission et al., 2013; TEEB, 2010). They are essential for human well-being by providing provisioning (e.g. food, timber, freshwater), regulating (e.g. climate regulation, water purification, soil erosion prevention) and cultural (e.g. recreation, aesthetic) services. Global change has been threatening and affecting many ecosystems, thereby causing severe modifications in ES and related benefits for human well-being (Karp et al., 2015; Mooney et al., 2009; Nelson et al., 2013; Runting et al., 2017). It is important to monitor changes in ES to ensure future sustainable development of our planet, but also to identify the drivers of change and related spatial and temporal scales at which they operate (Rounsevell et al., 2010). A driver of change is defined as a direct or indirect ecological or human-caused factor affecting ecosystem processes, functioning, and eventually ES (Millennium Ecosystem Assessment, 2005). Direct drivers of change have an explicit effect on ecosystems such as land use and land cover change, climate change, pollution, and the occurrence of invasive species. Indirect drivers are more diffuse, such as demographic, economic, sociopolitical, cultural, scientific, and technological drivers. Since both direct and indirect drivers can act at different spatial and temporal scales, they complicate the monitoring of their respective impact on ES.

Climate change with its associated effects on temperature, precipitation and sea level rise is among the most frequently investigated direct drivers of dynamics in ES supply (Runting et al., 2017; Scholes, 2016). The impact of climate change on ES is predominantly negative and varies cross ES, drivers and assessment methods with stronger negative impacts on regulating and cultural ES compared to provisioning ES (Runting et al., 2017). Particularly the spatial variability of climatological drivers causes difficulties in the assessment and management of their impacts on ES. Further, the impact of ES change drivers is often complex with multiple drivers affecting state or condition of ecosystems. This complexity determines a changing capacity of ecosystems which, in combination with altered ecosystem use (e.g. through adaptation), eventually leads to dynamics in ES supply (e.g. Hein et al., 2015). Besides direct drivers, there are also indirect drivers of change due to

socioeconomic pressures (e.g. population growth, demand for agricultural commodities and pricing) that act simultaneously and can cause non-climatological drivers, for example land cover and land use change. Land cover and land use change have been one of the most frequently studied non-climatological drivers with often negative impacts on ES (Egarter Vigl et al., 2017; Lawler et al., 2014; Polasky et al., 2011; Runting et al., 2017).

Investigating the influence of climatological and non-climatological drivers simultaneously is essential to understand their complex interactions, and their relative and cumulative impact on ES change. Although an increasing number of studies assess the influence of ES change drivers, the focus has mainly been on selected individual drivers and their impact on ES supply as well as on predictions of ES under future climate and land use change scenarios (Runting et al., 2017). The assessment of multiple drivers and multiple ES is often neglected, only a few studies focus on several socioecological indicators to explain spatial variations in multiple ES (Meacham et al., 2016; Mouchet et al., 2017). However, it is known that mutual interactions of several simultaneous drivers on ES need to be considered to fully understand ES changes (Runting et al., 2017). Disentangling the contribution of all drivers including their individual and cumulative impact on ES alterations is a remaining problem. Furthermore, most known causal relationships between ES and change drivers are based on future projections of ES dynamics that only consider land use and climate change scenarios (Lawler et al., 2014; Martinez-Harms et al., 2017; Nelson et al., 2010; Rosenzweig et al., 2014). Still, few studies evaluate the effect of change drivers on actually measured ES trends (Egarter Vigl et al., 2016; Guerra et al., 2016b; Nelson et al., 2013; Schirpke et al., 2013). This determines a gap since identifying the actual impact of change drivers on ES allows evaluating the effect and efficiency of past decisions and can improve decision making in landscape planning and conservation. We consequently hypothesize that monitoring approaches investigating actual ES and underlying change drivers affecting ES are essential. Such approaches allow facilitating environmental schemes that incorporate ES in their agenda, e.g. to monitor the efficiency of payments for provided ES, to assess the progress towards political targets regarding ES restoration (e.g. EU Biodiversity Strategy 2020, the Aichi

Biodiversity Targets and the Sustainable Development Goals), and to reevaluate those measures if necessary.

The aim of this study is to investigate possibilities to disentangle spatially the effect of climatological and non-climatological drivers on ES supply and trends. We explored time series of three ES in Switzerland between 2004 and 2014, namely CO₂ regulation, soil erosion prevention, and air quality regulation. We used an additive model to describe the spatial variation attributed to changes in climatological drivers (i.e. temperature, precipitation, and relative sunshine duration) and non-climatological random effects representing other spatially structured processes possibly affecting ES change. We investigated non-climatological effects at regional scale to reveal their underlying causes (e.g. land use change due to political, economic or environmental decision-making).

We realize that 11 years is a relatively short period to examine the effects of climate change on ES with climate variability affecting trends in climatic factors relatively strongly at such a short time frame. Nevertheless, we believe our approach is useful for studying the impacts of climatic trends on ES – even though these effects are likely to be more pronounced when studied at longer time frames.

2. Methods

2.1 Study area

Switzerland is located in central Europe with an area of approximately 4.1 million ha. The country covers altitudes ranging from 196 m to 4634 m above sea level (a.s.l.) in four biogeographical regions: (1) the Jura, a low mountain range in the northwest, (2) the Swiss Midlands, lowlands crossing Switzerland from the west to the northwest, and the (3) Central and (4) Southern Alps, a mountain range covering about 60 % of the country (Figure 1).

Land use and land cover are characterized by urban and settled areas covering 7.5 % of the country, agriculture (35.8 %), wooded areas (31.3%), surface water (4.3 %), and remaining natural environment (21.0 %) (Swiss Federal Statistical Office, 2016). During the last 25 years, the Swiss

landscape has constantly changed due to urban expansion particularly in the Swiss Midlands (Jaeger and Schwick, 2014) and due to

land abandonment in Alpine areas followed by expansion of wooded areas (Gellrich et al., 2007).

2.2 Data

2.2.1 Ecosystem service time series

We investigated time series of three ES, namely CO₂ regulation (CO₂R), soil erosion prevention (SEP), and air quality regulation (AQR) (Table 1) between 2004 and 2014 with annual time increments. The investigation period is limited due to the availability of all input data. The calculation of ES is outlined hereafter, more details on the ES models and data used can be found in Braun et al. (2018). We used linear regression analysis to derive ES trend maps with a 1 km spatial resolution and resampled them to 2 km (bilinear interpolation) to match the spatial resolution of investigated climatological variables (cf. Appendix S1: Figure S2 a-c for significant trends).

Note that the recent IPBES classification followed a different typology and uses the concept of "nature's contributions to people" instead of ecosystem services (see Díaz et al., 2015). However, we maintain the term ES in order to be aligned with SEEA, CICES as well as our earlier work (Braun et al., 2018).

 CO_2 regulation (CO_2R)

 CO_2R represents the role of vegetation in mitigating climate change and was defined as net ecosystem production (*NEP*) without considering harvest losses due to a lack of temporally and spatially explicit harvest data. NEP (g C m⁻² yr⁻¹) was estimated according to Schröter et al. (2014) as the difference of annual net primary production (*NPP*) (g C m⁻² yr⁻¹) and annual soil respiration (R_S) (g C m⁻² yr⁻¹) as:

$$CO_2R = NEP = NPP - R_S$$
, with (1)

$$R_S = \sum 1.250 \cdot e^{(0.05452 \cdot T_a)} \cdot \left(\frac{P}{4.259 + P}\right). \tag{2}$$

Annual NPP was derived from the annual MODIS NPP (MOD17A3H) product (Running and Zhao, 2015). Annual R_S was calculated according to Raich et al. (2002) as yearly sum (Σ) of mean monthly air temperature (T_a) (°C) and mean monthly precipitation (P) (mm) provided in 2 km spatial resolution (MeteoSwiss, 2013). Braun et al. (2017) showed in their landscape scale study that estimates of R_S and associated NEP partly deviate from flux tower observation for certain land cover types (e.g. grassland), due to the double accounting of below ground respiration and the coarse spatial resolution of used monthly meteorological data.

Air quality regulation (AQR)

AQR describes the ability of vegetation to filter aerosols from the atmosphere. It was calculated as yearly sum (Σ) of integrated daily measurements representing the vertical capture of particulate matter of less than 10 μ m (PM₁₀) by ecosystems (Manes et al., 2016; Nowak, 1994) as

$$AQR = \sum C \cdot V_d \cdot LAI \cdot T \cdot 0.5 \tag{3}$$

C is the PM₁₀ concentration in the air (μ g m⁻³) and was provided by the Swiss Federal Office for Environment (2015), V_d is the dry deposition velocity for PM₁₀ (m s⁻¹), a constant reported in Remme et al. (2014), LAI is the leaf area index (m² m⁻²) represented by the MODIS LAI product (MOD15A2H) (Myneni et al., 2015), T is the time step (here corresponding to one year), and 0.5 is the suspension rate of deposited PM₁₀ returning back to the atmosphere (Zinke, 1967).

Soil erosion prevention (SEP)

We defined SEP according to Guerra et al. (2016a) as the difference between the structural impact Y (e.g. the total soil erosion impact in the absence of soil erosion prevention = potential soil erosion)

and the remaining ES mitigated impact β (i.e. the remaining soil erosion that was not regulated by soil erosion prevention) as:

$$SEP = \Upsilon - \beta = \Upsilon (1 - C_{V}), \text{ with}$$
 (4)

$$Y = R \cdot K \cdot LS,\tag{5}$$

where Y is the structural impact (t ha⁻¹), C_V is the vegetation cover (-), R is the rainfall erosivity (MJ mm ha⁻¹ h⁻¹) (Panagos et al., 2015), K is the soil erodibility (t ha h ha⁻¹ MJ⁻¹ mm⁻¹) (Panagos et al., 2014), and LS describes the effect of topography on soil erosion (-), estimated with the terrain analysis module of the SAGA (System for Automatic Geoscientific Analysis) software (Pilesjö and Hasan, 2014). Vegetation cover C_V was estimated as a function of NDVI as suggested by Guerra et al. (2016a):

$$C_V = \exp\left[-2 \cdot \frac{NDVI}{(1-NDVI)}\right]. \tag{6}$$

The MODIS NDVI product (MOD13A1) (Didan, 2015) was used to calculate the vegetation cover C_V . Further details about this ES estimation can be found in Guerra et al. (2016a) and in Braun et al. (2018).

The calculation of ES incorporates three MODIS products (i.e. NDVI, LAI, NPP) related to vegetation density. However, retrieval concepts underlying these MODIS products substantially differ and the products represent different plant trait categories including structural (LAI), a combination of structural and biochemical (NDVI), and functional (NPP) traits. Further, the combination of MODIS products with environmental variables using linear and non-linear functions to obtain ES resulted in a highly non-linear influence of underlying MODIS products across retrieved ES.

2.2.2 Climatological data

MeteoSwiss provided annual mean temperature, precipitation, and relative sunshine duration data from 2004 to 2014 with a spatial resolution of 2 km (MeteoSwiss, 2013). The annual temperature (version 1.4) and precipitation data (version 1.0) were obtained from spatial interpolation of in situ measurements of about 80 weather stations and 420 rain gauges in Switzerland (MeteoSwiss, 2013). Annual relative sunshine duration data (version 1.2) were based on a combination of in situ heliometer measurements of about 70 stations and high-resolution satellite imageries (MeteoSwiss, 2013). We calculated trends for all three climatological drivers by linear regression (Appendix S1 Figure S1 and Figure S2 d-f for significant trends).

Changes in climatological drivers can cause alterations in ecosystem properties, functioning and eventually ES. We particularly selected temperature, precipitation and sunshine duration assuming that they show complex positive or negative effects on ES (Figure 2). For example, all three climatological drivers negatively influence SEP, since changes in them have negative effects on vegetation growth. AQR shows highly complex dependencies on climatological drivers (Jacob and Winner, 2009). Increasing precipitation, for example, can have a positive effect on air quality itself (reduction of PM₁₀) but might cause a negative effect on AQR: AQR is dependent on PM₁₀ concentrations, means less PM₁₀ in the atmosphere implies that less PM₁₀ can be filtered by vegetation, causing a reduced provisioning of AQR. Similar as AQR, CO₂R shows complex dependencies on investigated climatological drivers while a change of these drivers can possibly have a positive or negative effect on CO₂R: Less organic biomass will be decomposed due to less precipitation resulting in less CO₂ losses due to heterotrophic respiration and an increase in CO₂R. However, vegetation might reduce its photosynthetic activity and, thus, NPP due to limited water availability, resulting in a decrease in CO₂R.

2.3 Determination of climatological and non-climatological effects

We used an additive model to describe observed temporal changes in ES (y, response). The model contained a deterministic part g(x), where y depends on a set of covariates x with their coefficients, a spatial process h representing all non-climatological drivers that may influence ES and a residual component ε (de Jong et al., 2013):

$$y = g(x) + h + \varepsilon \tag{7}$$

We adapted this model by using a random forest (R package "randomForest", version 4.6, Liaw and Wiener, 2002) as deterministic part to model the influence of climatological drivers on ES change. We refer to the following modeled deterministic component as *DC*. The relationship between ES and climatological drivers may depend on biogeographical regions because of their distinctly different nature, e.g. mountains versus midlands, which render simple linear models inadequate. For this, we split the data set into training (70 %) and test (30 %) parts and set the number of trees to 1000. The contribution of the individual climatological drivers to the overall modeled deterministic component of each ES was estimated by calculating the importance of the variable based on mean decrease in accuracy using the R package "caret" (Kuhn, 2015, version 6.071). The direction of the relationship between climatological drivers and the explained variation in ecosystem services was based on the sign of the respective correlation coefficients.

The spatial process was estimated from the remaining spatial variance of observed changes in ES and the modeled deterministic component y-DC and may be attributed to effects by non-climatological drivers or to interactions that cannot be explained by the climatological drivers. A transformation of y-DC per ES was necessary to meet the stationarity assumption of the spatial process model (transformation CO_2R : log(y-DC); SEP: (y- $DC)^{0.25}$; AQR: (y- $DC)^{0.2}$). Before the transformation, the absolute minimum value per ES was added to account for negative values.

The related spatial process h was modeled by a Gaussian process having zero mean and a parameterized spherical covariance function (de Jong et al., 2013): a range parameter that represents the length of the spatial dependence and a sill parameter that describes the marginal variance. We This article is protected by copyright. All rights reserved.

term the variance of the white noise component as the nugget. The initial sill and nugget were estimated by method-of-moments using gstat sample variograms (Pebesma and Wesseling, 1998) and the range between 0.1 and 1.2 degrees (ca. 135 km) was tested using maximum likelihood estimation to assess the length of the spatial dependence. In order to account for the heterogeneous topography of Switzerland, we set the range to 1.2 degrees (ca. 135 km) for all ES. The residuals were calculated as the difference between the observations, the modeled deterministic component DC and the spatial process h (Eq. 7) after reversing the transformation.

The standardized ratio index between effects of climatological and non-climatological drivers (*RI*) indicates which type of driver was strongest in explaining the spatial variance in ES trends and was calculated as:

$$RI = -1 + \frac{2 \cdot |DC|}{|DC| + |h|} = \frac{|DC| - |h|}{|DC| + |h|} \tag{8}$$

Positive values of RI represented a stronger influence of climatological drivers with 1 being only affected by climatological drivers, while negative RI values represented influences of non-climatological drivers with -1 being completely without influence of climatological drivers. RI equals zero indicated the same contributions of effects by climatological and non-climatological drivers to ES change. Additionally, the influence of climatological drivers on ES change was estimated by calculating the coefficient of determination (R^2) of the observed ES change and the modeled deterministic component (DC) as quotient of the residual sum of squares and the total sum of squares:

$$R^{2} = \frac{\sum (y_{i} - DC_{i})^{2}}{\sum (y_{i} - \bar{y}_{i})^{2}},$$
(9)

where y_i are the values of observed ES change, DC_i are the predicted values of ES change due to effects by climatological drivers and \bar{y} is the mean value of the observed ES change.

3. Results

3.1 Effects of climatological drivers on ES change (deterministic component g(x))

The modeled deterministic component g(x) described the effect of climatological drivers on ES change. The spatial variation in ES change was well captured by the random forest models (Figure 3b) indicated by R^2 values of 0.63 for SEP followed by 0.54 for AQR and 0.40 for CO_2R (Table 2). At regional scale, the Jura and Swiss Midlands obtained the highest spatial variance explained by climatological drivers for all ES, while ES change in the Central Alps was less influenced by climatological drivers. All ES showed negative R^2 values in the Southern Alps and in the Swiss Midlands indicating that the deterministic model was not able to predict the effect of climatological drivers in the these regions for these ES. The individual climatological drivers contributed diversely to the modeled deterministic components of the different ES in Switzerland (Table 2). All three climatological drivers were nearly equally important for CO_2R and SEP. AQR was most affected by temperature that revealed the highest variable importance with 0.52. Furthermore, precipitation was approximately twice as important as relative sunshine duration for AQR.

The relationship between individual climatological drivers and the deterministic component revealed several negative, neutral, and positive directions (Figure 4). For instance, all climatological drivers indicated positive effects on SEP. In contrast, AQR revealed mainly negative effects of climatological drivers except for a neutral interaction of AQR with temperature. Furthermore, CO₂R was negatively influenced by precipitation, while temperature and relative sunshine duration revealed neutral effects.

3.2 Effects of non-climatological drivers on ES change (spatial process h)

The spatial process of the additive model described the spatial variation in ES trends that could not be attributed to climatological drivers (Figure 3c). These effects were smaller for all ES compared to the This article is protected by copyright. All rights reserved.

effect of climatological drivers. For all ES, the spatial variation of low negative and positive ES trends in the Jura and Swiss Midlands was attributed to the component representing non-climatological drivers. Also, stronger negative and positive ES trends in the Central and Southern Alps were captured by this non-climatological driver component. This also holds true for decreasing trends in the Southern Alps in CO₂R and AQR, and increasing trends in SEP in the southwestern Central Alps.

3.3 Ratio of effects by climatological and non-climatological drivers

The standardized ratio of effects by climatological and non-climatological drivers indicated the relative importance of both types of drivers to explain spatial variations of ES change. The results showed that ES trends were mainly attributed to changes in climatological drivers in Switzerland (positive values in Figure 3d). In particular, trends in SEP were strongly affected by climatological drivers with an *RI* value of 0.58 for Switzerland followed by CO₂R (*RI* 0.42) and AQR (*RI* 0.27). Regional differences revealed that climatological drivers explained most of the ES trends in the Jura and Swiss Midlands for all ES and the changes in SEP in the Central and Southern Alps. Effects of non-climatological drivers dominated only local changes in CO₂R and AQR in the Central and Southern Alps. However, these areas revealed also small-scale heterogeneity of climatological and non-climatological drivers. AQR showed a larger area with *RI* values around zero in the southwest of the Central Alps.

4. Discussion

4.1 Climatological drivers and ES change

In general, there has been relatively little land use change in Switzerland over the investigated period (European Environmental Agency, 2016; Swiss Federal Statistical Office, 2016) and, consequently, trends in all ES were mainly attributed to climatological drivers in Switzerland. Explained effects of climatological drivers varied across ES and regions with mainly negative, but also positive influences

on ES. The impact of climatological drivers over Switzerland explained well the negative ES trends in CO₂R and AQR, which is in line with other studies (Runting et al., 2017). Temperature was both positively and negatively correlated to CO₂R in Switzerland with some spatially distributed positive effects on increasing CO₂R trends in parts of the Central Alps. This positive effect of temperature on CO₂R in the Swiss Alps can be explained by forest expansion into higher altitudes (Gellrich et al., 2007). The widely distributed negative CO₂R trends across Switzerland could be equally attributed to changes in all three climatological drivers due to their similar variable importance. This indicates that temperature might have both positive and negative effects on CO₂R trends, although spatially separated: positive in the Central Alps, where increasing temperature can have a strong effect on vegetation growth in higher altitudes, and negative in the other parts of Switzerland, where increasing temperature might be accompanied with a deficiency in water availability. Lu et al. (2013) detected such varying effects of temperature on carbon sequestration for different ecosystem types such as forest, grassland, shrubland, and tundra. This explanation also applies to our analysis, since small-scale heterogeneity in land use and land cover with different ecosystem types, exposition and underlying environmental conditions characterizes Switzerland.

Trends in SEP were, on average, positively and equally affected by climatological drivers in Switzerland. This contradicts current findings of ES studies, where individual climatological drivers (i.e. temperature, precipitation and rel. sunshine duration) mainly revealed negative impacts on SEP and soil fertility (Runting et al., 2017). However, it seems that an increase in temperature (cf. Appendix S1 Figure S1a) in Switzerland between 2004 and 2014 fostered vegetation growth and consequently an increase in SEP. Particularly in the Southern Alps and parts of the Central Alps, the positive effect of temperature even compensated an expected decrease in SEP caused by increasing precipitation (Appendix S1 Figure S1b).

Furthermore, we attributed 54 % of the trends in AQR to climatological drivers, with temperature contributing strongest to observed changes. Particularly in the Jura and Swiss Midlands, effects of climatological drivers dominated trends in AQR, while effects of climatological and non-climatological drivers equally influenced AQR in the Central Alps. Observed dominance of

climatological drivers in the Jura and the Swiss Midlands can be partly explained by negative effects of precipitation on AQR (Appendix S1 Figure S1b). Negative effects of precipitation on AQR are also confirmed by another study investigating the impact of climate change attributes on particulate matter driven air quality (Jacob and Winner, 2009). Temperature revealed a neutral effect and relative sunshine duration a negative effect on AQR changes, which contradicts Jacob and Winner (2009). It must be, however, noted that results in Jacob and Winner (2009) are based on few studies only and that particularly the effect of temperature is uncertain since temperature estimates are often too inaccurate.

The heterogeneous spatial pattern in ES trends were well captured in the modeled deterministic components, despite a few local over- and underestimates mainly located in the Southern Alps. The Southern Alps seemed to represent a special case with similar topography and ES provisioning as for the Central Alps, but with differences in climatological drivers. This caused problems in the deterministic model g(x) resulting in local over- and underestimates. We acknowledge that our analysis regressed both long-term climate trends and inter-annual climate variability with ES supply. Our models convincingly showed that variability in climatological drivers affects the supply of ES from one year to the next, and that there is major spatial variability in this process. The investigated research period of 11 years was relatively short compared to what is needed to capture complex responses of ecosystems to climate change and related effects on ES. Hence, further research incorporating longer time series, is suggested to confirm how changes in ecosystems due to climate change affect the capacity of ecosystems in generating ES.

4.2 Non-climatological drivers and ES change

Effects of non-climatological drivers explain ES trends in Switzerland to a smaller magnitude compared to effects of climatological drivers. Particularly for CO₂R and AQR, effects of non-climatological drivers dominantly explained ES trends but only in spatially restricted areas in the Central Alps ES. This might be attributed to small-scale land cover and land use change in

Switzerland, which is in general almost negligible considering the whole area of the country in the investigated period (Swiss Federal Statistical Office, 2016). Indeed, changes in CORINE land cover maps between 2006 and 2012 revealed very few differences, while no hotspot areas of land cover change can be identified in Switzerland (European Environmental Agency, 2016).

Additionally, decreasing PM_{10} concentrations due to reduced PM_{10} emissions in Switzerland between 2004 and 2014, particularly in the southern part of Switzerland (Appendix S1 Figure S3), and a decline in natural and remote landscapes could cause the observed influence of non-climatological drivers on AQR trends. However, the detected influence of non-climatological drivers could be partly associated to artifacts in the spatial process (h) model compensating for possible overestimates of the deterministic component (g(x)). The deterministic component obtained higher trends in these areas compared to observed ES trends and represents an overestimate. Indeed, effects of non-climatological drivers seemed to cause observed increasing trends in AQR in the southwest of the Swiss Midlands and the Central Alps. However, trends of the deterministic component exceeded observed AQR trends in these areas, representing an artifact due to an overestimation of the modeled deterministic component.

Slightly dominating effects of non-climatological drivers caused a decrease in CO₂R in the Southern Alps. The effective non-climatological driver determining these negative trends is unclear; potential explanations are land use changes, an underestimation of the deterministic component resulting in artifacts in the explained variance of the non-climatological driver, and further processes that are unknown. Increasing CO₂R trends explained by effects of non-climatological driver might be the result of an underestimated deterministic component. Additionally, interactions between effects of climatological and non-climatological drivers, for example, of climate and land use change, might be another reason for the observed CO₂R trends.

Effects of non-climatological drivers also influenced SEP trends in the Central and Southern Alps and particularly explained spatial patches with relatively high SEP trends. Potential explanations for these SEP trends could be land cover changes, e.g. due to afforestation, and changes in ecosystem management that might have caused an increase in vegetation cover.

4.3 Reliability of the analysis

Applied remote sensing based time series of ES enabled detecting the impact of climatological and non-climatological drivers on trends in ES supply. Since the investigated time series was relatively short with only 11 years, we cannot rule out uncertainties in obtained trends of ES and climatological drivers due to anomalous years.

Modeled deterministic components estimating the contribution of climatological drivers on ES trends were based on random forest regression. The applied modelling accounted for non-linear effects in relationships between ES and climatological drivers. The interpretation of obtained results at regional scale indicated that the modeled deterministic component for the Southern Alps could not be estimated correctly using a random forest approach. Small-scale heterogeneity in trends of climatological drivers over all Switzerland caused prediction errors, because the deterministic model could not predict the resulting diversity of ES trends. Training and predicting the deterministic component per region and not at national scale could solve this problem (cf. Appendix S2). Consequently, explained effects of climatological and non-climatological drivers on ES trends should be carefully interpreted.

The deterministic model only considered climatological drivers but other drivers can be included too, for example land use change as one of the most frequently assessed non-climatological driver in ES studies (Egarter Vigl et al., 2016; Lawler et al., 2014; Nelson et al., 2010; Runting et al., 2017). The importance of land use change can be explained by its well-accepted relevance for ecosystem processes, the availability of land use change models, and found evidence on its negative impact on ES (Foley et al., 2005; Runting et al., 2017). Furthermore, many ES mapping approaches are built on

land cover data (Lavorel et al., 2017), fostering research on future ES changes based on land cover scenarios (Lawler et al., 2014; Martinez-Harms et al., 2017; Nelson et al., 2010; Nelson and Daily, 2010). The strong focus on land use change likely causes a bias by currently neglecting other possibly relevant drivers such as pollution and processes showing long-term dynamics (e.g. soil organic matter accumulation). Our analysis allowed disentangling effects of climatological from non-climatological drivers. We did not consider specific non-climatological drivers but our interpretation revealed a cumulative effect of non-climatological drivers beyond just the effect of land use change.

Analyzing climatological and non-climatological drivers and their effects on ES supply is critical for understanding the complexity involved in the drivers impacts on ES (Bryan, 2013; Carpenter et al., 2009). Our analysis contributes to this challenge by quantifying cumulative and individual contributions of climatological drivers to ES change, as well as the spatial structure of other non-climatological drivers. Only a few studies (Runting et al., 2017; Schirpke et al., 2017) combined both types of drivers in their analysis, while most focused exclusively on climate change (Nelson et al., 2013; Scholes, 2016) or land use change (Guerra et al., 2016b; Martinez-Harms et al., 2017; Nelson et al., 2010; Schirpke et al., 2013; Stürck et al., 2015). Therefore, our analysis represents a step forward in tackling the challenge of investigating effects of both climatological and non-climatological drivers on ES. Furthermore, the investigation of cumulative and individual impacts of climatological drivers on ES has been limited so far (Runting et al., 2017), although it provides further insights and information that can improve management decision in landscape planning and conservation. Moreover, this information is also crucial for evaluating the progress towards policy targets such as the Aichi Biodiversity Targets and the Sustainability Goals and to provide guidelines for implementing the ES concept efficiently into practice (Cord et al., 2017).

5 Conclusions and outlook

We demonstrated that an approach building on additive models offers new pathways to disentangle effects of climatological and non-climatological drivers on ES trends. We conclude that particularly the application of a random forest model as deterministic component allows compensating limitations of linear models by accounting for spatially heterogeneous ES trends as, for example, present in Switzerland. However, the random forest model is scale dependent and future work should consider adapting it for subscale analysis.

Our findings revealed that effects of climatological drivers on ES change in Switzerland are stronger compared to non-climatological drivers. In general, climatological and non-climatological drivers affected ES both negatively and positively, varying across the regions, drivers, and services assessed. In Switzerland, climatological drivers explained most of the trends in SEP followed by AQR and CO₂R. All three climatological drivers equally influenced trends in SEP and CO₂R, while temperature followed by precipitation mainly drove trends in AQR.

ES trends were mainly affected by climatological drivers in lower altitudinal areas (Jura and Swiss Midlands), and more strongly affected by non-climatological drivers in higher altitudes in the Central Alps. We consequently conclude on the importance of conducting regional ES monitoring and an associated analysis of the effects of relevant change drivers.

We suggest future research to extend the impact assessments of change drivers on biophysical ES supply to investigations of ES flow and demand. Additionally, we identify the need for advanced assessments considering the combined effects of climatological and non-climatological drivers on ES to systematically identify affected ES and their spatial locations and extent.

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DATA AVAILABILITY

Data are available on the Zurich Open Repository and Archive (ZORA): [data DOI cannot be assigned until this paper's DOI is available, and thus will be provided at proof]]

Table 1. Overview of investigated ecosystem services (ES) and their input of data (cf. Braun et al, 2018 for details on the used methodology).

ES	ES potential supply	Input data	Spatial resolution (m)
CO ₂ regulation (CO ₂ R)	Sequestered CO ₂ [g C m ⁻² yr ⁻¹]	MODIS: land cover (MCD12Q1)	500
		Net primary production (MOD17A3H)	500
		Mean monthly air temperature	2000
		Mean monthly precipitation	2000
Air quality regulation (AQR)	PM ₁₀ removal [μg m ⁻²]	MODIS: Leaf area index (LAI) (MOD15A2H)	500
		Land cover (MCD12Q1	500
		Annual PM ₁₀ concentrations	200
		Dry deposition velocity for PM ₁₀	-
Soil erosion prevention (SEP)	Potential soil erosion [t ha ⁻¹]	MODIS: NDVI (MOD12A1)	500
		Digital elevation model	25
		Rainfall erosivity	500
		Soil erodibility	500

Table 2. Proportion of variance in ES trends explained by climatological drivers (deterministic component) (R^2) in Switzerland and its regions and variable importance of the individual climatological drivers (i.e. temperature, precipitation, rel. sunshine duration) to predict the modeled deterministic component in Switzerland.

Ecosystem service	CO ₂ regulation	Soil erosion	Air quality regulation
\mathbb{R}^2			
Jura	0.67	-0.62	0.71
Swiss Midlands	0.68	-0.10	0.71
Central Alps	0.39	0.48	0.51
Southern Alps	-0.52	-0.17	-0.11
Switzerland	0.40	0.63	0.54
Variable importance			
Temperature	0.34	0.35	0.52
Precipitation	0.31	0.31	0.32
Relative sunshine duration	0.34	0.34	0.16

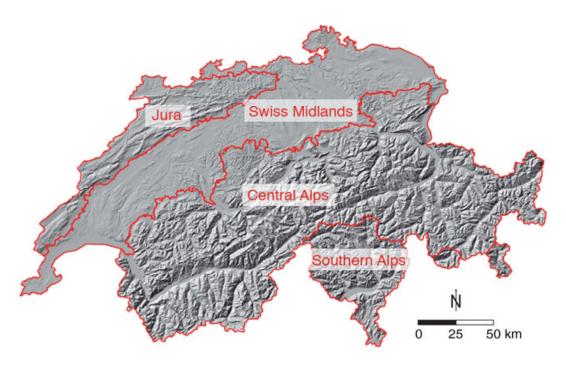
Figure legends

Figure 1 The study area Switzerland with its different geographical regions. A digital elevation model is used as background (courtesy of Swisstopo).

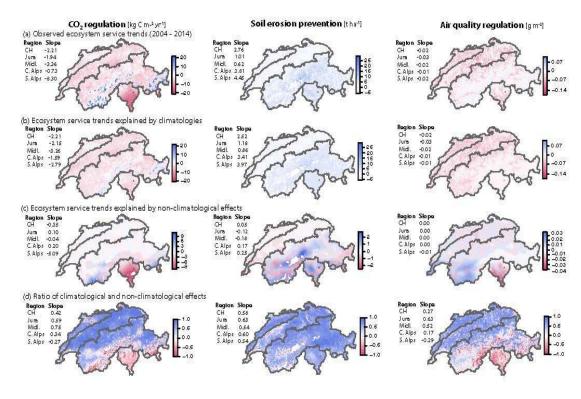
Figure 2 The impact of climatological drivers on ecosystem services (ES) (based on Runting et al. (2017) and Jacob and Winner (2009)). The influence of climatological drivers on ES can be positive (+, blue color), negative (-, red color) or mixed (±, green color), when studies found both positive and negative effects. The color represents the certainty level of the causal relationships and was estimated from several studies (number of studies are added in brackets below the sign). The darker the color the higher is the certainty level.

Figure 3 Observed ecosystem service (ES) trends between 2004 and 2014 (a), explained by effects of climatological drivers (b) and non-climatological drivers (c). The standardized ratio of effects by climatological and non-climatological drivers (d) represents the influence of (b) (positive, blue) and (c) (negative, red) on ES change. The tables contain average trends for Switzerland and its regions. Note that for visualization purposes the scales of (c) differ from (a) and (b).

Figure 4 Direction of the relationship between climatological drivers and the explained variation in ecosystem services based on the sign of the respective correlation coefficients. Red colors represent negative and blue colors positive correlations. The darker the color the stronger is the correlation.



	CO ₂ regulation	Soil erosion prevention	Air quality regulation
Temperature	- ± (16- 30)	- (6-15)	- (1)
Precipitation ^	± (6-15)	- (6-15)	- (1)
Precipitation $oldsymbol{\Psi}$	± (16-30)	- ± (6- 15)	+ (1)
Rel. sunshine duration	± (2-5)	- (1)	+ (1)



Climatology	CO ₂ regulation	Soil erosion prevention	Air quality regulation
Temperature	0.07	0.30	-0.09
Precipitation	-0.14	0.49	-0.28
Rel. sunshine duration	0.04	0.42	-0.22