Temperature signal in suspended sediment export from an Alpine catchment

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Abstract

Suspended sediment export from large Alpine catchments (> 1000 km²) over decadal timescales is sensitive to a number of factors, including long–term variations in climate, the activation–deactivation of different sediment sources (proglacial areas, hillslopes, etc.), transport through the fluvial system, and potential anthropogenic impacts on the sediment flux (e.g. through impoundments and flow regulation). Here, we report on a marked increase in suspended sediment concentrations observed near the outlet of the upper Rhône River Basin in the mid–1980s. This increase coincides with a statistically significant step–like increase in basin–wide mean air temperature. We explore the possible explanations of the suspended sediment rise in terms of changes in water discharge (transport capacity), and the activation of different potential sources of fine sediment (sediment supply) in the catchment by hydroclimatic forcing. Time series of precipitation and temperature–driven snowmelt, snow cover and icemelt simulated with a spatially distributed degree–day model, together with erosive rainfall on snow–free surfaces, are tested to explore possible reasons for the rise in suspended sediment concentration. We show that the abrupt change in air temperature reduced snow cover and the contribution of snowmelt, and enhanced ice–melt. The results of statistical tests show that the onset of increased icemelt was likely to play a dominant role in the suspended sediment concentration rise in the mid–1980s. Temperature–driven enhanced melting of glaciers, which cover about 10% of the catchment surface, can increase suspended sediment yields through increased contribution of sediment–rich glacial meltwater, increased sediment availability due to glacier recession, and increased runoff from sediment–rich proglacial areas. The reduced extent and duration of snow cover in the catchment are also potential contributors to the rise in suspended sediment concentration through hillslope erosion by rainfall on snow–free surfaces, and increased meltwater production on snow–free glacier surfaces. Despite the rise in air temperature, changes in mean discharge in the mid–1980s were not statistically significant, and their interpretation is complicated by hydropower reservoir management and the flushing operations at intakes. Overall, the results show that to explain changes in suspended sediment transport from large Alpine catchments it is necessary to include an understanding of the multitude of sediment sources involved together with the hydroclimatic conditioning of their activation (e.g. changes in precipitation, runoff, air temperature). In addition, this study points out that climate signals in suspended sediment dynamics may be visible even in highly regulated and human impacted systems. This is particularly relevant for quantifying climate change and hydropower impacts on streamflow and sediment budgets in Alpine catchments.

1. Introduction

Erosion processes and sediment dynamics in Alpine catchments are determined by geological, climatic and anthropogenic, factors. Geological forcing is one of the main drivers of sediment production and landscape development, through crustal thickening, deformation and isostatic uplift, and glacier inheritance (e.g. England and Molnar, 1990; Schlunegger and
Glacier inheritance influences sediment production and transport as demonstrated by a strong spatial association between sediment yield and past and current glacial cover (Hinderer et al., 2013; Delunel et al., 2014). Almost continuous temperature–driven glacier recession in the European Alps since the late 19th Century (Paul, 2004; 2007; Haeberli, 2007) has maintained large parts of the landscape in early stages of the paraglacial phase where unstable or metastable sediment sources (Ballantyne, 2002; Hornung et al., 2010) can maintain high sediment supply rates. Anthropogenic impacts on sediment yields are more recent, and on a global scale largely related to land cover change through intensified agriculture and the trapping of sediment in reservoirs (e.g. Syvitski et al., 2005). Land use changes impact mainly fine sediment production (e.g. Foster et al. 2003; Wick et al., 2003), while river channelization, flow regulation, water abstraction, and sediment extraction have caused a general reduction in sediment yield and consequently led to sediment–starved rivers world–wide (Kondolf et al., 2014). In Alpine catchments, in addition to trapping in reservoirs, sediment transfer is also disturbed by flow abstraction at hydropower intakes. The reduction of sediment transport capacity downstream of intakes and the periodic flushing of locally trapped sediment has severe impacts on the sediment budget (e.g. Anselmetti et al., 2007) and downstream river ecology (e.g. Gabbdud and Lane, 2016).

Here we focus on the dominant role of climate in sediment production and transfer in Alpine environments (e.g. Huggel et al., 2012; Zerathe et al., 2014; Micheletti et al., 2015; Palazon and Navas, 2016; Wood et al., 2016). The premise behind this work is that to explain impacts of changes in climate on Alpine catchment suspended sediment yield, it is necessary to consider both transport capacity and sediment supply. Sediment supply depends on many factors, most importantly the spatial location of sediment sources (e.g. lithology, distance to outlet, connectivity) and the specific processes of sediment production (e.g. hillslope erosion, glacial erosion, release of subglacially stored sediment, channel bed and bank erosion, mass wasting events) and transport (e.g., hysteresis).

In this study we look at specific sediment sources and the hydroclimatic conditioning of their activation (e.g. precipitation, runoff and air temperature) with a process–based perspective with the aim to infer the possible effects of changes in hydroclimate, such as increases in temperature and/or precipitation intensity, on suspended sediment dynamics. We identify four main sediment sources typical of Alpine environments: glacial erosion, hillslope erosion, channel bed/bank erosion and mass wasting events (e.g. rockfalls, debris flows). Climatic conditions, specifically precipitation and air temperature, contribute to the activation of these four sediment sources through different processes and at different rates. Erosive processes of abrasion, bed–rock fracturing and plucking at the base of glaciers provide proglacial areas with large amounts of sediment (Boulton, 1974). Due to glacial erosion, discharge from subglacial channels has high suspended sediment concentrations (e.g. Aas and Bogen, 1988). Temperature–driven snow and icemelt in spring and summer, as well as intense rainfall on snow–free surfaces, may lead to entrainment from proglacial areas provided they are connected to the river network (Lane et al., 2017). Hillslope erosion driven by overland flow and rainfall erosivity may be exacerbated in Alpine catchments by permanently or partially frozen ground (Quinton and Carey, 2008). In summer and autumn, when Alpine catchments are largely free of snow, intense rainfall may erode large amounts of sediment and transport it in rills and gullies to the river network. Intense rainfall is also responsible for triggering mass wasting events, such as debris flows and landslides, where a large mass of sediment is delivered to the channel network instantaneously (e.g. Bennett et al., 2012).

Flow conditions (e.g. shear stress, stream power) then determine the sediment transport capacity and in–stream sediment mobilization along rivers, and hence its transfer to downstream locations.

The close link between precipitation, air temperature, runoff and the activation–deactivation of sediment sources in Alpine catchments becomes critical in the context of climate change. Alpine regions represent a sensitive environment in relation to current rapid warming. In Switzerland, together with glacier recession, a reduction in snow cover duration and mean snow depth has been observed during the last thirty years (e.g. Beniston, 1997; Laternser and Schneebeli, 2003; Scherrer et al., 2004; Marty, 2008; Scherrer et al., 2006). Although current effects of climate change are less clear for precipitation
(Brönnimann et al., 2014) than for temperature, a sharp reduction in the number of snowfall days has been observed at many meteorological stations in Switzerland (Serquet et al. 2011).

The upper Rhône River Basin draining into Lake Geneva in Switzerland is at the center of our investigation. The basin has experienced a rise in air temperature that coincided with a rise in suspended sediment concentrations in the mid–1980s. Our main objective is to explore the presence of the signal of a warmer climate in the suspended sediment dynamics of this regulated and human–impacted Alpine catchment. In this work, we refer to fine sediment as the sediment transported in suspension. To investigate the potential causes of the observed increase in suspended sediment concentration, we conceptualize the upper Rhône Basin as a series of spatially distributed sediment sources that are activated or deactivated by hydroclimatic forcing. In addition to discharge (transport capacity), we consider four main hydroclimatic variables: (a) icemelt runoff (IM) which evacuates accumulated fine sediment, the product of glacial erosion, through subglacial channels (e.g. Swift et al., 2005); (b) snow cover fraction (SCF) which influences icemelting onset, impacts icemelt efficiency through albedo and may result in more rapid erosion and sediment production through an increased glacier basal velocity (e.g. Herman et al., 2015); (c) snowmelt runoff (SM) from snow–covered areas which may generate downstream hillslope erosion and channel erosion (e.g. Lenz et al., 2003); and (d) effective rainfall (ER), defined as liquid precipitation over snow–free areas, which lead to hillslope erosion, mass wasting and, due to enhanced discharge, also channel erosion (e.g. Bennet et al., 2012; Meusburger and Alewell, 2014). Our aims are: (a) to estimate daily basin–wide icemelt, snow cover fraction, snowmelt and effective rainfall over the Rhône Basin for the last 40 years; and (b) to analyze these variables with the goal to provide statistical evidence for possible reasons for the rise in suspended sediment concentrations in the mid–1980s.

2. Study Site Description

The upper Rhône Basin is located in the southwestern part of Switzerland, in the Central Swiss Alps (Fig. 1). It has a total surface area of 5338 km², and an altitudinal range of 372 to 4 634 m a.s.l. About 10% of the surface is covered by glaciers, which are mostly located in the eastern and southeastern part of the catchment (Stutenbecker et al., 2016). The Rhône Glacier originates at the Rhône Glacier and flows for about 160 km through the Rhône valley before entering Lake Geneva, a few kilometers downstream of the gauging station located at la Porte–du–Scex (Fig. 1). Basin–wide mean annual precipitation is about 1400 mm yr⁻¹ and shows strong spatial variability driven mostly by orography and the orientation of the main valley. The hydrological regime of the catchment, typical of Alpine environments, is strongly influenced by snow and icemelt with highest discharge in summer and lowest in winter. Mean annual discharge is 180 m³ s⁻¹, which corresponds to about 1060 mm yr⁻¹ and an annual runoff coefficient of 75%.

The catchment has been strongly affected by anthropogenic impacts during the last century. The main course of the Rhône River has been extensively channelized for the purposes of flood protection: levees were constructed and the channel was narrowed and deepened in the periods 1863–1894 and 1930–1960 (First and Second Rhône Corrections). Due to the residual flood risk that affects the main valley, a third project was started in 2009 with the main objectives to increase channel conveyance capacity and river ecological rehabilitation (Oliver et al., 2009). In addition, significant gravel mining operations are carried out along the main channel and many tributaries. Since the 1960s, several large hydropower dams have been built in the main tributaries of the Rhône River. The total storage capacity of these reservoirs corresponds to about 20% of the mean annual streamflow (Loizeau and Dominik, 2000). Flow impoundment, water abstraction and diversion through complex networks of intakes, tunnels and pumping stations, have significantly impacted the flow and sediment regime of the catchment. Flow regulation due to hydropower production has resulted in a considerable decrease of discharge in summer and increase in winter (Loizeau and Dominik, 2000). That said, the construction of dams and start of hydropower operation has coincided with a drop in the suspended sediment load of the main Rhône River measured at la Porte–du–Scex in the 1960s (Loizeau et al., 1997; Loizeau and Dominik, 2000).
Two sub-catchments of the Upper Rhône basin are used for the calibration/validation of the icemelt model: the Massa and the Lonz (Fig. 1). The Massa is a medium-sized basin (195 km²) with a mean elevation of 2945 m a.s.l. More than 60% of the surface is glacierized, and the remaining surface is classified mostly as rock and firm (Boscarello et al., 2014). The basin includes the Aletsch Glacier, which is the largest glacier in the European Alps with a length (1973) of around 23.2 km and a surface area (1973) of approximately 86 km² (Haebeli and Holzhauser, 2003). The Lonz is a relatively small basin located to the west of the Massa with an average elevation of 2630 m a.s.l. It has a total drainage area of roughly 77.8 km² and almost 36% of its surface (1991) is covered by glaciers.

3. Methods

Our objective is to explore the potential effect of climate on suspended sediment dynamics of the upper Rhône Basin in the period 1975–2015. To this end, we analyze observed and simulated hydroclimatic and sediment transport variables, as listed in Table 1: mean daily temperature $T$, total daily precipitation $P$, mean daily discharge $Q$, suspended sediment concentration SSC, daily snow cover fraction SCF, icemelt SM, icemelt IM, and effective rainfall over snow–free areas (ER). Some variables originate from observations ($Q$, SSC) or spatial interpolation of observations ($T$, $P$), others from simulations by spatially distributed snow and icemelt models (SCF, SM, IM), or from a combination thereof (ER). The snowmelt model is described in Sect. 3.1, the icemelt model in Sect. 3.2, and their calibration in Sect. 3.3. We first interpolate the input datasets of precipitation and temperature on a 250×250 m grid resolution by the nearest-neighbor interpolation method. Second we run the snow and icemelt model at the daily timescale over the period 1975–2015. In a third step, we analyze the variables (Table 1) as mean monthly and mean annual values averaged over the basin area. To quantify changes in the hydroclimatic variables and in suspended sediment concentration, we apply standard statistical tests for change detection described in Sect. 3.4. A description of all datasets used in this analysis is reported in more details in Sect. 4.

3.1 Snowmelt Model

We use a snowmelt model to predict SM and SCF over the entire basin, because snow station measurements are sparsely and irregularly distributed and a physical consistency between precipitation and air temperature as climatic driving forces and snowmelt and snow cover as response variables are needed. The spatially distributed temperature index method (degree–day model) was used due to its simplicity, low data requirements, and demonstrated success at daily temporal scales over large basins (e.g. Hock, 2003; Boscarello et al., 2014). The degree–day approach also matches the coarse spatial (250×250 m) and temporal (daily) resolution of our analysis and the areal averaging at the basin scale. Models based on energy balance, or enhancements of the degree–day approach, represent physical processes better and could be used when higher spatial and temporal resolution and accuracy is needed (e.g. Pellicciotti et al., 2005).

The snowmelt model includes snow accumulation and melt. At the grid scale, precipitation $P$ [mm day⁻¹] is first partitioned into solid and liquid form based on (a) daily minimum $T_{\text{min}}$ [°C] and maximum air temperature $T_{\text{max}}$ [°C] and (b) a rain–snow threshold temperature $T_{\text{RS}}$ [°C]. If minimum air temperature $T_{\text{min}}$ is above the threshold temperature $T_{\text{RS}}$, all precipitation falls as rainfall $R$; if the maximum air temperature $T_{\text{max}}$ is below the threshold temperature $T_{\text{RS}}$, all precipitation falls as snow $S$; otherwise precipitation is a mixture of liquid and solid form, partitioned proportionally to the temperature difference:

$$\begin{align*}
R &= c_p P \\
S &= (1 - c_p) P
\end{align*}$$

where

$$c_p = \frac{T_{\text{RS}} - T_{\text{min}}}{T_{\text{max}} - T_{\text{min}}}$$

1
\[
\begin{align*}
&c_p = 1 & T_{\text{min}} > T_{\text{RS}} \\
&c_p = 0 & T_{\text{max}} \leq T_{\text{RS}} \\
&c_p = \frac{T_{\text{max}} - T_{\text{RS}}}{T_{\text{max}} - T_{\text{min}}} & T_{\text{min}} \leq T_{\text{RS}} < T_{\text{max}}.
\end{align*}
\]

The daily snowmelt rate \( SM_i \) [mm day\(^{-1}\)] is estimated from a linear relation with air temperature:

\[
\begin{align*}
&SM_i = k_{\text{snow}}(T_{\text{mean}} - T_{\text{SM}}) & T_{\text{mean}} > T_{\text{SM}} \\
&SM_i = 0 & T_{\text{mean}} \leq T_{\text{SM}}.
\end{align*}
\]

where \( T_{\text{mean}} \) [°C] is the mean daily air temperature, \( T_{\text{SM}} \) [°C] is a threshold temperature for the onset of melt, and \( k_{\text{snow}} \) is a melt factor [mm day\(^{-1}\) °C\(^{-1}\)]. Snow depth (SD), in mm snow water equivalent, for time \( t \) is then simulated from a balance between accumulation and melt at every grid cell \( i \):

\[
SD_i(t) = SD_i(t - 1) + S_i(t) - SM_i(t).
\]

The snow cover fraction SCF for a chosen area containing \( i = 1, \ldots, N \) grids, is:

\[
\text{SCF}(t) = \frac{1}{N} \sum_{i=1}^{N} H[SD_i(t)].
\]

To reduce degrees of freedom, the threshold temperature for the onset of melt \( T_{\text{SM}} \) is set equal to 0°C, which is a typical value for Alpine regions (e.g. Schaefl et al. 2005; Corbari et al., 2009; Boscarello et al., 2014). The calibration of the snowmelt model consists of estimating only the melt factor \( k_{\text{snow}} \) with methods described in Sect. 3.3. In addition, we apply a sensitivity analysis on the three parameters \( k_{\text{snow}}, T_{\text{RS}}, T_{\text{SM}} \) as described in the Supplementary Material, Sect. S1.

### 3.2 Icemelt Model

Similar to snowmelt, icemelt is also simulated with a temperature index (degree–day) model on grid cells that are identified as glacier–covered. The daily icemelt \( IM_i \) [mm day\(^{-1}\)] on glacier surfaces that are snow–free is estimated as:

\[
\begin{align*}
&IM_i = k_{\text{ice}}(T_{\text{mean}} - T_{\text{IM}}) & T_{\text{mean}} > T_{\text{IM}} \\
&IM_i = 0 & T_{\text{mean}} \leq T_{\text{IM}}.
\end{align*}
\]

where \( T_{\text{mean}} \) [°C] is mean daily air temperature, \( T_{\text{IM}} \) [°C] is a threshold temperature for the onset of icemelt, and \( k_{\text{ice}} \) [mm day\(^{-1}\) °C\(^{-1}\)] is the icemelt factor. For the entire catchment, we estimate mean daily icemelt IM [mm day\(^{-1}\)] as the arithmetic average over all ice–covered grid cells:

\[
IM(t) = \frac{1}{N} \sum_{i=1}^{N} IM_i(t).
\]

The threshold temperature for glacier melting \( T_{\text{IM}} \) is set equal to 0°C. Icemelt occurs only if the glacier cell is snow–free. The snow cover simulated by the snowmelt model in Sect. 3.1 is thus essential for estimating icemelt. The calibration of the icemelt model consists of estimating only the melt factor \( k_{\text{ice}} \) as described in Sect. 3.3.
3.3 Calibration and Validation of Snowmelt and Icemelt Model

We perform the calibration and validation of the snow and icemelt model parameters in sequence, since the snow–covered surface is required for icemelt estimation on glaciers. The snowmelt factor \( k_{\text{snow}} \) is calibrated based on comparisons with snow cover maps. Snow cover observations are split into two periods: 1 October 2000 – 30 September 2005 for calibration and 1 October 2005 – 31 December 2008 for validation, for a total number of 217 calibration and 143 validation days. Snow cover maps at 500x500 m resolution are distributed by proximal interpolation to the snowmelt model 250x250m computational grid. Maps of snow depth simulated with Eq. (4) are first transformed into simulated snow cover fraction \( \text{SCF}^{\text{sim}} \) with Eq. (5) and are afterwards compared with snow cover fraction \( \text{SCF}^{\text{obs}} \) derived from the observations.

The objective function for calibration is based on a combination of mean absolute error and true skill statistic. The mean absolute error \( \text{MAE} \) is estimated as:

\[
\text{MAE} = \frac{1}{n} \sum_{j=1}^{n} \left| \text{SCF}^{\text{obs}}_{j} - \text{SCF}^{\text{sim}}_{j} \right|
\]

(9)

where \( n \) is the number of MODIS image maps. MAE captures the overall ability of the model to reproduce the snow cover fraction accurately. The true skill statistic \( \text{TSS} \) is a spatial statistic that measures the grid–to–grid performance of the model in capturing snow–no snow presence. It is computed as the sum of sensitivity \( \text{SE} \) (correct snow predictions) and specificity \( \text{SP} \) (correct no–snow predictions) computed from contingency tables (e.g. Wilks, 1995; Mason and Graham, 1999; Corbari, 2009) in each image \( j \) and averaged over the \( n \) MODIS maps in the simulation period:

\[
\text{TSS} = \frac{1}{n} \sum_{j=1}^{n} (\text{TSS}_{j}) = \frac{1}{n} \sum_{j=1}^{n} \text{SE}_{j} + \text{SP}_{j} - 1
\]

(10)

Because \( \text{TSS} \) includes both sensitivity and specificity, it captures both predictions of snow–covered and snow–free areas. It takes on values between 0 and 1, where 1 indicates perfect performance, and is a widely applied metric for assessing spatial model performance (e.g. Begueria, 2006; Allouche et al., 2006). We combine both goodness–of–fit measures (MAE and \( \text{TSS} \)) into an objective function \( \text{OF} \), by giving more weight to MAE. Finally, we evaluate the objective function \( \text{OF} \) over \( b = 5 \) different elevation bands in order to better capture the topographic gradients in snowmelt distribution in the Rhône Basin:

\[
\text{OF} = \sum_{b=1}^{5} \text{OF}_{b} = \sum_{b=1}^{5} -0.6 \text{MAE}_{b} + 0.4 \text{TSS}_{b}.
\]

(11)

This objective function is maximized in calibration. The rationale of using both MAE and \( \text{TSS} \) in evaluating performance is to give weight to both basin–integrated snow cover as well as to grid–based predictions. Indeed, the same value of snow cover fraction can result in two different spatial arrangements of snow–covered pixels, and a correct spatial distribution of snow–covered and snow–free areas is relevant for this analysis insofar as it affects the activation and deactivation of specific sediment sources. The weights assigned to MAE and \( \text{TSS} \) in Eq. (11) are the outcome of sensitivity tests with the model.

After calibration, we also estimate the Nash–Sutcliffe efficiency \( \text{NS} \) (Nash and Sutcliffe, 1970) and the mean square error \( \text{MSE} \) to quantify the performance of the model:

\[
\text{NS} = 1 - \frac{\sum_{j=1}^{n} (\text{SCF}^{\text{obs}}_{j} - \text{SCF}^{\text{sim}}_{j})^2}{\sum_{j=1}^{n} (\text{SCF}^{\text{obs}}_{j} - \text{SCF}_{j})^2}
\]

(12)

\[
\text{MSE} = \frac{1}{n} \sum_{j=1}^{n} (\text{SCF}^{\text{obs}}_{j} - \text{SCF}^{\text{sim}}_{j})^2
\]

(13)

where \( \text{SCF} \) is the average observed snow cover fraction during the calibration–validation period.

We calibrate the icemelt factor \( k_{\text{ice}} \) on the sub–basin of the river Massa (Fig. 1), on the basis of daily discharge measurements, and focusing only on months when the icemelt contribution is not negligible (June–October). The gauging station is located upstream of the Gebidem dam, therefore discharge is not influenced by reservoir regulation and represents undisturbed natural flow. Calibration is performed on the period 1 January 1975 – 31 December 2005, while validation covers the remaining ten years of available data, i.e. the period 1 January 2006 – 31 December 2015. We then validate the model on the Lonza sub–basin with the same procedures and goodness of fit measures.

The optimal value of \( k_{\text{ice}} \) is found by minimizing the mass balance error \( \text{MBE}_{e} \) computed for the period June–October:
\[
MBE_i = 100 \frac{\sum_{j=1}^{ny} (V_j^{\text{obs}} - V_j^{\text{sim}})}{\sum_{j=1}^{ny} V_j^{\text{obs}}} .
\]

where \(ny\) is the number of calibration years, \(V_j^{\text{obs}}\) and \(V_j^{\text{sim}}\) [mm year\(^{-1}\)] are the observed and simulated discharge volumes per unit area reaching the outlet of the catchment during the period June–October of each calibration year \(i:\)

\[
V^{\text{obs}} = \sum_{j=1}^{nd} Q_j^{\text{obs}},
\]

\[
V^{\text{sim}} = \sum_{j=1}^{nd} Q_j^{\text{sim}} = \sum_{j=1}^{nd} (R_j + SM_j + IM_j).
\]

Here, \(nd\) is the number of observation days from June to October, \(Q_j^{\text{obs}}\) [mm day\(^{-1}\)] is the daily discharge per unit area observed at Blatten Bei Naters (Blatten), \(R_j, SM_j, IM_j\) are respectively the total daily rainfall, snowmelt and icemelt aggregated over the Massa (Lonzar) basin. Rainfall (\(R\)) and snowmelt (\(SM\)) are simulated with the snow accumulation and melt model in Sect. 3.1, while icemelt (\(IM\)) is simulated with the icemelt model in Sect. 3.2.

It should be noted that in this study we do not consider glacier evolution, i.e. changes in ice thickness due to accumulation and melt, as well as glacier ice flow. Neglecting glacier retreat rises the possibility that we overestimate the icemelt contribution over the study period. To quantify the potential effect of glacier retreat, we compare our simulations with time series produced from the Global Glacier Evolution Model (GloGEM), a model accounting for both the mass balance and glacier evolution (Huss and Hock, 2015). For comparison, we use total monthly runoff generated from glacierized surfaces of the upper Rhône basin, simulated with GloGEM for the period 1980–2010. GloGEM computes the mass balance for every 10–m elevation band of each glacier, by estimating snow accumulation, snow and ice melt, and refreezing of rain and melt water. The response of glaciers to changes in mass balance is modelled on the basis of an empirical equation between ice thickness changes and normalized elevation range parametrized as proposed by Huss (2010). Normalized surface elevation changes \(\Delta h_s\) are derived for each elevation band from mass balance changes (mass conservation). Starting from initial values derived by the method of Huss and Farinotti (2012), ice thickness is updated at the end of each hydrological year by applying the relation between normalized elevation range \(h_r\) and normalized surface elevation change \(\Delta h_s\). The area of each glacier is finally adjusted by a parabolic cross–sectional shape of the glacier bed (Huss and Hock, 2015). GloGEM is calibrated and validated over the period 1980–2010 with estimates of glacier mass changes by Gardner et al. (2013) and in situ measurements provided by the World Glacier Monitoring Service.

3.4 Statistical Testing for Change

We use the non–parametric Pettit test (Pettitt, 1979) for the detection of the time of change (year–of–change) in the air temperature data. We then test the other variables (SSC, P, Q, SM, SCF, IM and ER) for changes in the mean (and variance) by splitting the time series into two periods before and after the identified year–of–change, and by applying two–sample two–sided \(t\)-tests for the equality of the means (and variances). The null hypothesis of no change is tested at the 5% significance level. The \(t\)-test is a parametric test commonly used in hydrology to assess the validity of the null hypothesis of two samples having equal means and unknown unequal variances. We apply the \(t\)-test to all hydroclimatic variables averaged at the annual and monthly timescales with the same year–of–change to determine which hydroclimatic variables, and therefore the activation or deactivation of which sediment sources, are possibly responsible for the observed changes in suspended sediment concentration.

In our catchment, SSC is sampled intermittently (twice per week). This might have an effect on the change detection analysis of the hydroclimatic variables. We estimate this potential effect by considering the hydroclimatic variables SM, IM, ER and Q only on days corresponding to SSC–measurement days. We compare these new time series with the original ones by estimating the cumulative distribution functions of the variables and by testing changes of mean monthly and annual values over time. We consider only the positive (non–zero) part of the distributions. Results are reported in Sect. 5.4.
4. Data Description

4.1 Precipitation and Air Temperature

For precipitation and air temperature we use spatially distributed datasets provided by the Swiss Federal Office of Meteorology and Climatology (MeteoSwiss). Total daily precipitation, mean, minimum and maximum daily air temperature are available on a \(2 \times 2\) km resolution grid for Switzerland (MeteoSwiss, 2013). All four datasets are developed by spatial interpolation of quality–checked data collected at MeteoSwiss meteorological stations (Frei et al., 2006; Frei, 2014). We apply the statistical analysis of change to basin–averaged values of precipitation and temperature and not to individual grid point values, which might be potentially affected by substantial interpolation errors. Moreover, the variability in time of the number of stations involved in the spatial interpolation may induce non–homogeneities in the datasets. This is particularly relevant when analyzing long–term changes as in the case of this study. Therefore, we verify the effects of potential non–homogeneities by using an experimental dataset developed by MeteoSwiss specifically for this research, based on a constant number of stations (294 for precipitation and 48 for temperature) for the period 1971–2013. We applied the statistical tests for detecting changes both on the original and the experimental datasets of P and T. Results of the statistical tests on the two datasets coincide. This confirms that temporally variable number of meteorological stations employed to build the product does not influence the changes detected in the original dataset.

4.2 Discharge and Suspended Sediment Concentration

We use daily discharge data measured by the Swiss Federal Office for the Environment (FOEN) at three gauging stations: la Porte–du–Scex (available since 1905), Blatten Bei Naters (available since 1931) and Blatten (available since 1956) (Fig. 1). For suspended sediment concentration, two in–stream samples per week collected by FOEN at la Porte–du–Scex are available since October 1964 (Grasso et al., 2012).

In this work, we focus on sediment transported in suspension. Previous analysis on the grain size distribution of suspended sediment at the outlet of the upper Rhône River reports a bimodal distribution, with mode diameters equal to 13.7 \(\mu\)m (silt) for the finer fraction and 39.6 \(\mu\)m (silt) for the coarser grains (Santiago et al., 1992). The composition of grains cover a wide range of values, including clay (16.9%), silt (64.7%) and sand (18.4%). The mean suspended sediment size is reported to be equal to 17.7 \(\mu\)m (silt), and the largest grains transported in suspension during summer high flow conditions are in the range of coarse sand (> 500 \(\mu\)m) (Santiago et al., 1992).

4.3 Snow Cover and Glacier Data

We use snow cover maps derived from satellite imagery for the upper Rhône Basin over the period 2000–2008 processed in previous research (Faticchi et al., 2015). We use the 8–day snow cover product MOD10A2 retrieved from the Moderate Resolution Imaging Spectroradiometer (MODIS) (Dedieu et al., 2010) for the calibration and validation of the snowmelt model. MOD10A2 is provided at a 500\times500\ m spatial resolution, where cells are classified as snow–covered, snow–free, inland water or cloud–covered. In order to reduce the impacts of clouds in estimating snow cover fraction, maps with cloud cover greater than 30% are excluded from the dataset, resulting in a total number of usable images equal to 360, i.e. on the average 40 days per year.

The surface covered by glaciers is assigned based on the GLIMS (Global Land Ice Measurements from Space) Glacier Database (Fig. 1). Ice–covered cells identified based on the GLIMS data of 1991 give more than 10% of the upper Rhône Basin as covered by ice with a total glacier surface of almost 620 km\(^2\).
4.4 Digital Terrain Model

We use a digital terrain model (DTM) with 250x250 m resolution (85409 cells in total, Fig. 1), obtained by resampling a finer model (25x25 m) provided by SwissTopo in the ETH geodata portal (GeoVITe). The DTM is used as a mask for extracting climatic inputs and for elevation information in the snowmelt modelling.

5. Results

5.1 Calibration of Snowmelt and Icemelt Models

The snowmelt factor, calibrated following the procedure described in Sect. 3.3, is $k_{\text{snow}} = 3.6 \text{ mm day}^{-1} \text{ °C}^{-1}$. The snowmelt model reproduces well the seasonal fluctuations of snow cover fraction (SCF) in the basin, with Nash–Sutcliffe efficiencies (NS) close to 0.90 and low mean square errors (MSE). The model maintains good performances also in the validation period showing slight reduction in the goodness of fit measures (Table 2). The temporal variability of SCF is also well simulated at the basin scale. Although the comparison between observed and simulated SCF is affected by the discontinuous nature of the MODIS data (8–day resolution), Fig. 2 shows that the model with a single constant $k_{\text{snow}}$ for the entire catchment reproduces the snow cover dynamics reasonably well for all of the studied elevation bands. At lower elevations, the model tends to slightly underestimate SCF in autumn and overestimate it in winter. The model performs better at higher elevation bands, even at the very highest elevations with permanent snow cover (Fig. 2 bottom). The spatial distribution of snow cover is satisfactory, with average values of sensitivity and specificity greater than 0.7 (Table 2). Goodness of fit measures indicate that, on average, more than 70% of snow–covered and snow–free pixels are correctly identified. The true skill score, which combines both metrics, results in values around 0.5 (Table 2). Snow cover duration maps averaged over the period 2000–2008 for MODIS observations and simulations show a good spatial coherence (Fig. 3). In summary, we conclude that the snowmelt model represents the spatial and temporal dynamics of snow cover in the Rhône Basin satisfactorily. Results of the sensitivity analysis on $k_{\text{snow}}, T_{RS}, T_{SM}$ are reported in the Supplementary Material (Sect. S1).

The icemelt factor, calibrated following the procedure described in Sect. 3.3, is $k_{\text{ice}} = 6.1 \text{ mm day}^{-1} \text{ °C}^{-1}$. Calibration and validation results are summarized in Table 2. In Figure 4 we show the seasonal pattern of basin–averaged IM, SM, and R simulated with the calibrated snow and icemelt model, together with discharge Q observed at the outlet of the two highly glacierized sub-catchments Massa (Fig. 4a) and Lonza (Fig. 4b). The fit of the simulated (computed as IM+SM+R) to the observed discharge is good, with mass balance errors about 7% for the Massa and 8% for the Lonza.

Although in our hydrological model we do not include glacier evolution, the annual runoff volumes (SM+IM+R) from glacierized areas during the period 1980–2010 correlates well with the results of GloGEM (Fig. 5a). Measures of performance confirm the agreement between the two models: the correlation coefficient is equal to 0.86 and the Nash–Sutcliffe efficiency is equal to 0.67. We also capture quite well the seasonal pattern of runoff generated from glacierized areas (Fig. 5b). Perhaps most importantly, GloGEM simulations show that total annual runoff is increasing throughout the period and there is no evidence for decreasing icemelt rates. This confirms that, although glaciers of the upper Rhône basin are retreating, melt–water discharges from glacierized and proglacial areas are increasing during the 1980–2010 period. As expected, total runoff from glacierized surfaces and icemelt is highly correlated (Fig. 5a, correlation coefficient = 0.95), thus indicating that the increase in total runoff is due to an increase of the icemelt component. Indeed, non–parametric Mann–Kendall tests indicate an increasing trend with 5% significant level. Trend slopes, estimated with the Theil–Sen estimator, confirm the agreement between the two models: we find a total runoff of ~27.65 mio m$^3$ year$^{-2}$ with GloGEM and ~21.71 mio m$^3$ year$^{-2}$ with our model. We also computed the basin–averaged mass balance accounting for snow accumulation and snow and icemelt for each hydrological year. The mean mass balance over the period 1980–2010 is equal to $-0.78 \pm 0.22$ m w.e. year$^{-1}$ which is within the uncertainty range of recent studies (Fischer et al., 2015). In summary:
although we do not account for glacier retreat, our model results agree well with state–of the art glaciological models that include glacier evolution. Both comparisons with GloGEM and our basin–averaged mass balance indicate that we are not significantly overestimating icemelt during the period 1975–2015.

5.2 Temperature, Precipitation, Discharge and SSC in the Rhône Basin

Mean annual air temperature shows a clear and statistically significant increase in 1987 (p–value < 0.01). A two–sample t–test for equal means (p–value < 0.01) confirms an increase in mean daily temperature greater than 1 °C (Fig. 6a). Statistical tests on monthly means reveal that the 1987 temperature jump is mainly in spring and summer months from March to August, while changes in the autumn and winter months are not statistically significant (Fig. 7a). For the period March–August, mean monthly temperatures have risen by about 1.2 °C on the average.

The change in air temperature around 1987 coincides with statistically significant changes in mean annual suspended sediment concentration (Fig. 6c). After the abrupt warming, mean annual suspended sediment concentrations are roughly 40% larger than before: average values have risen from 172 ± 6.86 mg l⁻¹ before 1987 up to 242 ± 14.45 mg l⁻¹ after 1987, where the ranges express the standard error of the mean. This change can be ascribed to statistically significant (p–value < 0.01) increases in summer (July–August) concentrations (Fig 7c). Suspended sediment concentration is also characterized by much larger inter–annual variability after 1987 than before: the standard deviation of mean annual SSC increases from ~ 32 mg l⁻¹ before 1987 up to ~ 78 mg l⁻¹ after (Fig. 6c). A statistically significant increase in the variance is confirmed with a two–sample F–test at 5% significant level.

While the upper Rhône Basin underwent an abrupt warming around 1987, mean annual precipitation (Fig. 6b) and mean monthly precipitation (Fig. 7b) do not change significantly in time. Likewise, mean annual discharge does not show any statistically significant change in 1987 (Fig. 6d). Mean monthly discharge (Fig. 7d) is characterized by a small statistically significant increase in winter (November–February) runoff, most likely due to increased snowmelt and possibly changes in hydropower generation.

5.3 Hydroclimatic Activation of Sediment Sources

Mean annual simulated snowmelt (SM) shows a decreasing tendency during the last thirty years (Fig. 8a). The reduction in snowmelt after 1987 occurs mostly in summer and early autumn (Fig. 9a) mainly due to poor snow cover (Fig. 9b). However, except July and September, the changes in all months are within the 95% confidence interval. The increase of snowmelt in March and April is due to warmer temperatures in spring. Results are coherent with the temporal evolution of simulated snow cover fraction, which is also gradually decreasing (Fig. 8b), especially in spring and summer (Fig. 9b). Statistical analysis reveals a step–like reduction of more than 10% for mean annual values of snow cover fraction in 1987 (p–value < 0.01).

Although mean annual and monthly precipitation were shown not to change significantly in the mid–1980s, effective rainfall (ER) on snow–free areas has increased, especially in early summer (Fig. 8d, Fig. 9d). Effective rainfall increases in conjunction with decreases in snow cover fraction, and a statistically significant jump is identified in 1987 (p–value < 0.01) (Fig. 8d). However, although snow cover fraction is significantly lower throughout the entire melting season, only June and especially July show statistically significant increases in ER after 1987 (Fig. 9d).

Our results show that the temporal evolution of icemelt is consistent with suspended sediment concentration rise. Although the change is rather gradual at the annual scale (Fig. 8c), the step–like increase in icemelt is evident in the ice–melting season (May–September) and reaches highest magnitudes in July and August (Fig. 9c) in conjunction with rises in suspended sediment concentration in those months (Fig. 7c).
The simultaneous increase in icemelt and decrease in snowmelt suggests that the abrupt warming has led to important alterations of the hydrological regime. To quantify this alteration, we compute the relative contribution of rainfall, snow and icemelt on the sum of these three components in July and August. The average relative contribution of icemelt has almost doubled after 1987 (from ~12% to ~22%, Fig. 10), while the relative contribution of snowmelt has reduced by more than 30% (from ~52.5% to ~35%, Fig. 10). This indicates the substantial effect of the sharp temperature rise on the basin hydrology.

5.4 Effect of Intermittent SSC sampling

The empirical cumulative distribution functions of total daily basin–averaged SM, IM, ER and Q, computed on all days (“all non–zero days”) and only on days corresponding to SSC–measurements (“SSC–measurement non–zero days”) is shown in Figure 11. Although extremely high and low values may indeed be missed by the non-continuous sampling, cumulative distributions of SM, IM, ER and Q on “SSC–measurement non–zero days” and on “all non–zero days” are similar. This indicates that, although SSC is measured at a fixed interval, the sampling captures accurately the process variability. In addition, results of the statistical tests on mean monthly and mean annual values of all analyzed hydroclimatic variables are unchanged. We therefore conclude that our results are not significantly influenced by the discontinuous nature of the SSC sampling.

6. Discussion

6.1 Snowmelt and Icemelt Models

The value of the snowmelt factor \( k_{\text{snow}} \) (3.6 mm day\(^{-1}\) °C\(^{-1}\)) is in agreement with previous studies carried out in this region. In the upper Rhône Basin, Boscarello et al. (2014) found a snowmelt factor equal to 4.3 mm day\(^{-1}\) °C\(^{-1}\) based on previous studies on the Toce Basin in Italy (Corbari et al., 2009). Calibration of a semi–lumped conceptual model for the three tributary catchments of the upper Rhône Basin – Lonza, Drance and Rhône at Gletsch – led to snowmelt factors equal to 6.1, 4.5 and 6.6 mm day\(^{-1}\) °C\(^{-1}\), respectively (Schaeffli et al. 2005). Differences in \( k_{\text{snow}} \) between this and previous studies are attributable to different temporal resolution of models, lengths of calibration datasets, type and thresholds of precipitation partitioning, climatic inputs, threshold temperature for melt, and others. We highlight that the higher performance of the model in simulating snow cover at the highest elevation in our study, where most of the glaciers are located, is a prerequisite for successful icemelt estimation. The underestimation of SCF in autumn and the overestimation in winter at lower elevations are likely related to errors in partitioning precipitation into solid and liquid form. One of the main problems of degree–day models is related to their poor performance in reproducing the spatial distribution of snow accumulation and melt in complex topography. The temperature–index approach does not take into account features that affect melting, such as topographic slope, aspect, surface roughness and albedo (Pellicciotti et al., 2005). However, in our case, the spatial distribution of snow cover is satisfactory. Sensitivity and specificity are characterized by a strong seasonal signal. In summer, when a large part of the basin is snow–free, it is much easier for the model to capture snow–free pixels correctly than snow–covered pixels. In winter, when the basin is largely snow–covered, the situation is reversed. We account for this by computing the true skill score which combines both sensitivity and specificity into a better representation of overall model performance.

Despite the large regional and temporal variability that characterizes icemelt factors, comparison with previous studies confirms that the calibrated value (7.1 mm day\(^{-1}\) °C\(^{-1}\)) is reasonable for the Alpine environment (e.g. Schaeffli et al., 2005; Boscarello et al., 2014). A range from 5 to 20 mm day\(^{-1}\) °C\(^{-1}\) has been reported in the literature (e.g. Hock, 2003; Schaeffli,
2005). It should be noted that, when calibrating the icemelt factor, we neglect evaporation (evapotranspiration). However, evaporation plays indeed a secondary role in the long–term water balance in Alpine environments compared to precipitation and snowmelt (Braun et al., 1994; Huss et al., 2008), especially at high elevation such as in the case of the Massa and Lonza sub–catchments.

Considering that the aim of this study is to evaluate long–term changes in hydroclimatology and sediment dynamics of the upper Rhône Basin and not the short–term variability of icemelt at the daily scale, we consider the Snowmelt and Icemelt model performances as satisfactory. In addition, we show that although our model does not account for glacier retreat, it does not overestimate the icemelt contribution during the period 1975–2015. However, considering climate projections further into the future, and glaciers that continue to retreat, the issue of future icemelt contribution will need to be revised. Under climate change, even the largest glacier in the basin, the Aletsch Glacier, is expected to shrink at a rate where its icemelt contribution would start decreasing before 2050 (Farinotti et al., 2012; FOEN, 2012; Brönnimann et al., 2014).

6.2 Changes in Hydroclimatolology and SSC

Abrupt temperature jumps, such as the one we observed in the upper Rhône basin, rather than gradual changes in air temperature have been observed globally (e.g. Jones and Moberg, 2003; Rebetez and Reinhard, 2008). Observations indicate that Switzerland has experienced two main rapid warming periods in the past, with the 1940s and 1980s being the warmest decades of the last century (Beniston et al., 1994; Beniston and Rebetez, 1996). The simultaneous increase in temperature and suspended sediment concentration indicates that changes in climatic conditions may effectively impact sediment dynamics, especially in Alpine environments where temperature–driven processes, like snow and ice–melt, have a strong influence on the basin hydrology. The statistically significant change in the SSC variance supports the finding that processes related to fine sediment regime of the upper Rhône Basin have been altered by changing climatic conditions, resulting in greater concentrations and higher variability of suspended sediment reaching the outlet of the basin. Conversely, differences in precipitation before and after 1987 are within the 95% confidence interval and are not statistically significant. Differences in discharge are also not statistically significant except in winter, when the suspended sediment concentration does not show changes. Therefore, it is very unlikely that the abrupt increase in suspended sediment concentration around mid–1980s in July and August is caused by changes in mean precipitation and/or discharge.

6.3 Hydroclimatic Activation of Sediment Sources

Our simulations of snowcover and melt are in agreement with snow observations across Switzerland. The decreasing tendency in snow cover after mid or late 1980s has been demonstrated for the Swiss Alps (Beniston, 1997; Laternser and Schneebeli, 2003; Scherrer et al., 2004; Marty, 2008; Scherrer et al., 2006). Snow depth, number of snowfall days, and snow cover show similar patterns during the last century: a gradual increase until the early 1980s, interrupted in late 1950s and early 1970s, and a statistically significant decrease afterwards (Beniston, 1997; Laternser and Schneebeli, 2003). Previous analyses also state that the reduction in snow cover after mid–1980s is characterized more by an abrupt shift than by a gradual decrease (Marty, 2008), in agreement with our simulations. The reduction in snow cover duration, which is observed to be stronger at lower and mid altitudes than at higher elevations, is mainly the result of earlier snow melting in spring due to warmer temperatures (Beniston, 1997; Laternser and Schneebeli, 2003; Marty, 2008). Moreover, by analysing 76 meteorological stations in Switzerland, Serquet et al. (2011) demonstrated a sharp decline in snowfall days relative to precipitation days, both for winter and early spring, showing the impact of higher temperature on reduced snowfall, independently of variability in precipitation frequency and intensity. Therefore, despite the high complexity that characterizes snow dynamics in the Alps (Scherrer et al., 2006; 2013), the dominant effect of temperature rise on snow cover
decline after late 1980s has been clearly shown (Beniston, 1997; Marty, 2008; Serquet et al. 2011; Scherrer et al., 2004; 2006).

The increase in potentially erosive rainfall, is partially confirmed by recent observations. Rainfall erosivity, expressed by the R–factor of the Revised Soil Loss Equation (Wischmeier and Smith, 1978; Brown and Foster, 1987), computed on the basis of 10 min resolution precipitation data, was recently analysed for Switzerland. Although, the upper Rhône Basin together with the Eastern part of Switzerland was found to have relatively low rainfall erosivity (low R–factor) compared to the rest of the country due to a lower frequency of thunderstorms and convective events (Schmidt et al., 2016), there is evidence of an increasing trend for the R–factor from May to October during the last 22 years (1989–2010) (Meusburger et al., 2012).

This suggests that the increase in effective rainfall on snow–free surfaces may have contributed to suspended sediment concentration rise, through a combination of reduced snow cover fraction, increased rainfall–snowfall ratio and possible increases in rainfall intensity at sub–daily scale. However, simulations show a statistically significant jump in effective rainfall in June and July, while SSC is significantly larger in July and August. Therefore, we argue that erosive rainfall alone is unlikely to explain the abrupt jump in suspended sediment concentration observed around mid–1980s.

Enhanced icemelt is coherent with the observed acceleration of Alpine glacier retreat after mid–1980s. Ground–based and satellite observations, combined with mass balance analysis, reveal that current rates of glacier retreat are consistently greater than long–term averages (Paul, 2004; 2007; Haeberli, 2007). Estimations of glacier area reduction rates indicate a loss rate for the period 1985–1999, which is seven times greater than the decadal loss rate for the period 1850–1973 (Paul, 2004). Investigations with satellite data and in–situ observations suggest that the volume loss of Alpine glaciers during the last thirty years is more attributable to a remarkable down–wasting rather than to a dynamic response to changed climatic conditions (Paul, 2004; 2007). Haeberli et al. (2007) estimated that glaciers in the European Alps lost about half of their total volume (roughly 0.5% year\(^{-1}\)) between 1850–1975, another 25% (1% year\(^{-1}\)) between 1975–2000, and an additional 10–15% (2–3% year\(^{-1}\)) in the period 2001–2005. The appearance of proglacial lakes and rock outcrops with lower albedo and high thermal inertia, separation of glaciers from the accumulation area, and general albedo lowering in European Alps (Paul, 2005), are among the main positive feedbacks that accelerate glacier disintegration and make it unlikely to stop in the near future (Paul, 2007). Although glacier dynamics are quite complex and involve many variables and feedbacks, the dominant role played by temperature rise in glacier wasting has been clearly demonstrated (e.g. Oerlemans, 2000). The major volume loss in the recent past in Swiss Alpine glaciers is attributable to negative mass balances during the ablation season rather than to a lower accumulation by precipitation (Huss, 2008). For small high altitude Alpine glaciers, Micheletti and Lane (2016) showed negligible ice melt contributions to runoff between the mid–1960s and mid–1980s, after which contributions increased markedly.

Most importantly, runoff coming from glaciers is notoriously rich in sediments. Very fine silt–sized sediment resulting from glacier erosion is transported in suspension most often as wash load (Aas and Bogen, 1988). Proglacial areas generally represent rich sources of sediment due to active glacier erosive processes of abrasion, bed–rock fracturing and plucking (Boulton, 1974; Hallet et al., 1996). Glacier retreat discloses large amount of sediments available to be transported by proglacial streams. Moreover, change in climatic conditions and specifically temperature–driven glacier recession and permafrost wasting may initiate specific erosional processes that consequently enhance sediment supply in proglacial environments (Micheletti et al., 2015; Micheletti and Lane, 2016; Lane et al. 2017).

As shown in Sect. 5.3, icemelt increase is highest in July and August (Fig. 9c), in agreement with the jump in suspended sediment concentration (Fig. 7c), while ER rise occurred mainly in June and July (Fig. 9d). We then conclude that the significant increase in icemelt detected in the mid–1980s (Fig. 8c, 9c, 10) is likely to be the main cause of the sharp rise in suspended sediment concentration entering Lake Geneva, through a combination of: (1) increased discharge originated in proglacial environments, which implies higher suspended sediment concentration; (2) larger relative contribution of
sediment–rich icemelt compared to snowmelt and precipitation fluxes; and (3) intensified sediment production and augmented sediment supply in proglacial areas due to rapid ice recession.

6.4 Anthropogenic Factors and Climate Signals

The interpretation of increases in suspended sediment concentration may be complicated by anthropogenic drivers and changes in the mid-1980s. Three main anthropogenic activities may have potentially influenced the suspended sediment regime of the upper Rhône basin: river channelization, construction of reservoirs and hydropower operations, and gravel extraction along the main stream and tributaries. However, the second and last large channelization project was completed in 1960 (Oliver et al., 2009), much earlier than the observed increase in SSC. Likewise, the largest reservoirs in the catchment have been in operation since 1975 (Loizeau and Dominik, 2000). Therefore, it is unlikely that these two anthropogenic factors have contributed to the SSC rise detected in mid–1980s. The same holds for gravel mining activities. Annual volumes of gravel extracted from the Rhône, provided by the Cantonal Authorities as differences from the average over the period 1989–2014, do not show any significant correlation with mean annual suspended sediment concentration ($R^2 = 0.08$). Although gravel mining operation may perturb SSC for short periods after river bed disturbance by causing local pulses of fine sediments, this process does not affect significantly the suspended sediment load at the outlet of the basin over seasonal and annual timescales. A possibility still remains that changes in the hydropower operation itself, i.e. the distribution of flow responding to electricity demand, and the flushing of dams have increased SSC concentrations. We currently do not have any evidence for such changes, however we think it is unlikely that they would have long–term effects on SSC.

Our results show that even in highly human impacted and regulated catchments such as the Rhône Basin, a strong climatic signal in hydrological and sediment dynamics can persist. This also suggests that the decrease in fine sediment load at the outlet of the upper Rhône Basin observed in the 1960s on the basis of sediment cores recovered in the Rhône delta region and reported by Loizeau et al. (1997), could be the result of a combined effect of hydropower system development, as it has been hypothesized (Loizeau et al., 1997; Loizeau and Dominik, 2000), but also reduced icemelt loads due to colder temperatures at the time. The cooling period, which occurred between 1950s and late 1970s (e.g. Beniston et al., 1994) was characterized by colder and snowy winters (e.g. Laternser and Schneebeli, 2003) and has been accompanied by reduced icemelt rates, glacier advance and positive glacier mass balances (Zemp et al., 2008; VAW–ETH, 2015).

The climate signal in sediment dynamics takes on particular importance in the context of climate change projections into the future. Despite the large uncertainty, future projections under different climate change scenarios show a common tendency for Switzerland, characterized by a shift from snow–dominated to rain–dominated hydrological regime, reduced summer discharge, increased winter discharge, reduced snow cover, and enhanced glacier retreat (Bavay, 2009; Jouvet et al., 2011; Brönnimann et al., 2014; Fatichi et al., 2015; Huss, 2016). In contrast to these hydrological predictions, changes in sediment fluxes are highly uncertain due to the complexity and feedbacks of the processes involved, inherent stochasticity in sediment mobilization and transport, and large regional variability in sediment connectivity across the Alpine landscape (Cavalli et al., 2013; Heckmann and Schwanghart, 2013; Bracken et al., 2015; Lane et al., 2017).

7. Conclusions

The aim of this research was to analyze changes in the hydroclimatic and suspended sediment regimes of the upper Rhône Basin during the period 1975–2015. We show an abrupt increase in basin–wide mean air temperature in the mid–1980s. The simultaneous step–like increase in suspended sediment concentration at the outlet of the catchment, detected in July and August, suggests a causal link between fine sediment dynamics and climatic conditions. Two main factors link warmer climate and enhanced SSC: increased transport capacity and increased sediment supply resulting from spatial and/or temporal activation–deactivation of sediment sources. Our results show that transport capacity, through discharge, is not
likely to explain the increases in SSC, because no statistically significant changes in the mid–1980s are present in Rhône Basin discharge, neither at the annual nor monthly timescales. The suggestion is that the impact of warmer climatic conditions acts on fine sediment dynamics through the activation and deactivation of different sediment sources and different sediment production and transport processes.

5 To understand sediment supply conditions we analyze the temporal evolution of three main sediment fluxes: (1) sediments sourced and transported by snowmelt along hillslopes and channels; (2) sediments entrained and transported by erosive rainfall events over snow–free surfaces, including hillslope and channel bank erosion, and mass wasting events; (3) fine sediment fluxes generated by glacier ice–melt. The fluxes of snow and icemelt together with snow cover fraction and rainfall are analyzed to detect changes in time and their coherence with changes in SSC.

10 Our results show that while mean annual precipitation does not show any evident change between the periods before and after the SSC jump in mid–1980s, potentially erosive rainfall clearly increases over time especially in June and July, but not in August. On the other hand, icemelt has significantly increased due to temperature–driven enhanced ablation. Statistically significant shifts in icemelt were identified for summer, with highest increases in July and August, in accordance with the rise in SSC. Concurrently to the temperature and SSC rise, the relative contribution of icemelt to total annual runoff (sum of rainfall, snow and ice–melt) presents a significant increase in mid–1980s, substantially altering the hydrological regime of the Rhône Basin. Based on these results we propose that climate has an effect on fine sediment dynamics by altering the three main fluxes of suspended sediment in the Rhône Basin, and that icemelt plays a dominant role in the suspended sediment concentration rise in the mid–1980s through: (1) increased flow derived from sediment–rich subglacial and proglacial areas; (2) larger relative contribution of sediment–rich icemelt compared to snowmelt and precipitation; and (3) increased sediment supply in hydrologically connected proglacial areas due to glacier recession. While snowmelt has decreased, the reduced extent and duration of snow cover may also have contributed to the suspended sediment concentration rise through enhanced erosion by heavy rainfall events over snow–free surfaces. Because changes in SSC are not consistent with changes in discharge and transport capacity, our work emphasizes how the inclusion of sediment sources and their activation through different processes of production and transport is necessary for attributing change. This analysis also demonstrates that climate–driven changes of suspended sediment dynamics may be significantly strong even in highly regulated and human impacted catchments such as the Upper Rhône basin, where sediment fluxes are affected by flow regulation due to hydropower production and by grain–size dependent trapping in reservoirs. This has consequences for climate change impact assessments and projections for Alpine catchments with hydropower systems, where climate change signals are sometimes thought to be secondary to human regulation. Although at this stage we cannot reliably conclude in which direction sediment fluxes will change in the future, our paper clearly shows that a more process–based understanding of the connections between hydrological change and the activation of sediment sources will provide us with a better framework for analysing and attributing changes in sediment yields in Alpine catchments in the future.

Author contribution

35 A. Costa and P. Molnar designed the methodology. A. Costa developed the code and carried out simulations and computations. A. Costa prepared the manuscript with contributions from all co–authors. The authors declare that they have no conflict of interest.

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References


Table 1: List of the variables analysed: observed SSC, and hydroclimatic variables originating from measurements (T, P), from simulations of the snow and icemelt model (SCF, SM, IM), or a combination thereof (ER). Information on the source and the spatial and temporal resolution are reported for each variable.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Data Source</th>
<th>Resolution</th>
</tr>
</thead>
<tbody>
<tr>
<td>T</td>
<td>Daily mean temperature [°C] on a ~2×2 km grid provided by MeteoSwiss</td>
<td>basin–averaged, daily, 1975–2015</td>
</tr>
<tr>
<td>P</td>
<td>Daily total precipitation [mm day⁻¹] on a ~2×2 km grid provided by MeteoSwiss</td>
<td>basin–averaged, daily, 1975–2015</td>
</tr>
<tr>
<td>Q</td>
<td>Daily mean discharge [m³ s⁻¹] at three stations (Porte du Scex, Blatten, Blatten bei Naters) provided by FOEN</td>
<td>daily, 1975–2015</td>
</tr>
<tr>
<td>SSC</td>
<td>Suspended sediment concentration [mg l⁻¹] at la Porte du Scex provided by FOEN</td>
<td>2 times per week, 1975–2012</td>
</tr>
<tr>
<td>SCF</td>
<td>Snow cover fraction [0–1] simulated by the Snowmelt Model on a 250×250 m grid, and calibrated with MODIS satellite data for the period 2000–2009</td>
<td>basin–averaged, daily, 1975–2012</td>
</tr>
<tr>
<td>SM</td>
<td>Snowmelt rate [mm day⁻¹] simulated by the Snowmelt model on a 250×250 m grid</td>
<td>basin–averaged, daily, 1975–2012</td>
</tr>
<tr>
<td>IM</td>
<td>Icemelt rate [mm day⁻¹] simulated by the Icemelt model on a 250×250 m grid, and calibrated at Blatten and Blatten bei Naters</td>
<td>basin–averaged, daily, 1975–2012</td>
</tr>
<tr>
<td>ER</td>
<td>Effective rainfall [mm day⁻¹] (rainfall on snow–free pixels), estimated from P, T and SCF on a 250×250 m grid</td>
<td>basin–averaged, daily, 1975–2012</td>
</tr>
</tbody>
</table>

Table 2: (top) Calibrated snowmelt factor \( k_{\text{snow}} \) and goodness of fit measures for validation and calibration period: Nash–Sutcliffe efficiency (NS), mean square error (MSE), true skill statistic (TSS), sensitivity (SE) and specificity (SP) for the entire upper Rhône Basin; (bottom) calibrated icemelt factor \( k_{\text{ice}} \) and goodness of fit measures: mass balance error computed on June–October months (MBEₛ) and on the entire year (MBEₐ) for Massa and Lonza sub–basins.

\[
k_{\text{snow}} = 3.6 \text{ mm day}^{-1} \text{ °C}^{-1}
\]

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<td>SE</td>
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<td>SP</td>
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\[
k_{\text{ice}} = 6.1 \text{ mm day}^{-1} \text{ °C}^{-1}
\]

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<th>MBEₛ [%]</th>
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<td>Validation Lonza</td>
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Figure 1: Map of the upper Rhône Basin with topography, glacierized areas and river network. Inset shows the position of the upper Rhône Basin in Europe (blue). Locations of gauging stations used in this analysis are shown as triangles. Massa and Lonza sub–basins used in the calibration and validation of the ice melting component are highlighted.
Figure 2: Comparison between observed (red circles) and simulated (light blue lines) snow cover fraction (SCF) of the upper Rhône Basin for five different elevation bands. Simulations are computed with calibrated snowmelt factor $k_{\text{snow}} = 3.6$ mm day$^{-1}$ °C$^{-1}$. 
Figure 3: Map of average snow permanence over the period 2000–2008, expressed as fraction of time in which pixels are snow-covered (SCDF [0–1]): (a) observations (MODIS) and (b) simulations.

Figure 4: Comparison of mean monthly observed (dark blue) and simulated discharge (light blue) for the period 1975–2015: (a) Massa basin and (b) Lonza basin. Simulated discharge is the sum of three components: icemelt (IM), snowmelt (SM) and rainfall (R).

Figure 5: Runoff (snowmelt + icemelt + rainfall) generated from glacierized areas within the upper Rhône basin, simulated with GloGEM and with the snowmelt and icemelt models (degree–day) for the period 1980–2010: (a) total annual values; (b) mean monthly values. Fig. 5a also depicts the time series of total annual icemelt simulated with the icemelt model.
Figure 6: Observations for the period 1965–2015 of: (a) basin–averaged air temperature; (b) basin–averaged daily precipitation; (c) suspended sediment concentration measured at the outlet of the basin; (d) daily discharge per unit area measured at the outlet of the basin. Mean annual values are shown in grey and the 5–years moving average is shown with a bold line.

\[ \Delta T \approx 1.26 \, ^\circ C \]

\[ \Delta SSC \approx 70 \, \text{mg/l} \]
Figure 7: Monthly differences between the period after and before the year-of-change (1987–2015 and 1965–1986) of: (a) basin–averaged air temperature; (b) basin–averaged daily precipitation; (c) mean suspended sediment concentration measured at the outlet of the basin; (d) daily discharge per unit area measured at the outlet of the basin. Points outside the confidence interval (grey shaded area) represent statistically significant (5% significance level) changes in the monthly mean.
Figure 8: Simulations for the period 1975–2015 of mean annual: (a) snowmelt SM; (b) snow cover fraction SCF; (c) icemelt IM; (d) effective rainfall ER. Mean annual values are shown in grey and a 5–year moving average is shown with a thick line.
Figure 9: Mean monthly differences (between 1987–2015 and 1975–1986) in: (a) snowmelt SM; (b) snow cover fraction SCF; (c) icemelt IM; (d) effective rainfall ER. Points outside the confidence interval (grey shaded area) represent statistically significant (5% significance level) changes in the monthly mean.

Figure 10: Relative contribution of snowmelt (SM), rainfall (R), and icemelt (IM) for the summer months July–August, computed as the ratio between each component and their sum. Rainfall is extracted from observed precipitation by using a rain–snow temperature threshold, snow and icemelt are simulated with spatially distributed temperature–index models.
Figure 11: Empirical cumulative distribution functions of total daily basin-averaged SM (a), IM (b), ER (c) and Q (d), computed on all days and only on days corresponding to SSC–measurements. Only positive values of SM, IM and ER are included.
Supplementary Material: Temperature signal in suspended sediment export from an Alpine catchment

S1. Sensitivity Analysis on Snow Model parameters

We apply a sensitivity analysis on the three main parameters of our Snowmelt model: snowmelt factor $k_{\text{snow}}$ [mm day$^{-1}$ °C$^{-1}$], threshold temperature for the onset of snow melt $T_{\text{SM}}$ [°C], and the rain–snow threshold temperature $T_{\text{RS}}$ [°C]. We analyze the impact of the parameters on the model results by perturbing one single parameter at the time. We compute some of the goodness of fit measures adopted during the calibration (TSS, NS RMSE) and analyze their relative change as function of the parameter perturbation (Fig. S1).

As expected, the snowmelt factor $k_{\text{snow}}$ is the most sensitive parameter (Fig. S1). For $k_{\text{snow}}$ between 1.6 and 5.6 mm day$^{-1}$ °C$^{-1}$, the relative reduction of TSS and NS results respectively lower than 10% and 15% (Fig. S1). Although reducing the snowmelt factor $k_{\text{snow}}$ below 2.6 mm day$^{-1}$ °C$^{-1}$ increases RMSE by almost 40%, it results in incrementing RMSE only by 0.06 units (SCF [0–1]). For $T_{\text{SM}}$ varying within the range of $-2 \div 2$ °C, TSS and NS decrease less than 10%. While the change in RMSE results larger similarly to the case of $k_{\text{snow}}$ (Fig. S1). The effect of $T_{\text{RS}}$ is even more negligible, with relative changes of goodness of fit measures within 5.5% (Fig. S1).

We also estimated the influence of these three parameters on the trends and the jumps that we identified over the period 1975–2015 for the hydroclimatic variables simulated with the model: SM, IM, ER, SCF. For this purpose, we applied statistical tests for equality of mean annual and monthly values before and after mid–1980s after perturbing one parameter at the time within physically reasonable ranges of values.

In agreement with the results of our analysis (Sect. 5.3), a statistically significant increase of mean annual IM and ER and a statistically significant decrease of mean annual SCF after mid–1980s are detected for all selected parameter values. Similarly, mean annual SM shows a decreasing tendency after mid–1980s for all parameter selections, and in about 50% of all the parameters selections statistical tests even show a statistically significant drop in mean annual SM. The increases in mean monthly ER in June and July are identified respectively in more than 90% and in 100% of the cases. Similarly, all
parameters sets show an abrupt rise of IM after mid–1980s for all spring and summer months (May–August). This confirms that by perturbing the parameters of the snow model within reasonable ranges, the overall results of our analysis do not change. This is also depicted in Fig. S2. The differences in mean monthly values of IM and ER between the period after and before mid–1980s are shown for multiple values of each of the three parameters. The confidence interval depicted in Fig. S2 is built by selecting, for each month, the highest and the lowest values of the confidence interval (5% significance level) among all parameter sets. The comparison between Fig. S2 and Fig. 8c and Fig. 8d, shows that changing parameters within reasonable ranges would not substantially change the results of our analysis.

Figure S2: Sensitivity analysis on monthly differences between the periods after and before the year–of–change (1987–2015 and 1975–1986) for IM (left) and ER (right). Box plots represent monthly differences for all selections of parameters within given ranges of values: 1.6 mm day$^{-1}$ $\leq k_{\text{snow}} \leq 5.6$ mm day$^{-1}$ (a,b); $-2^\circ C \leq T_{\text{SM}} \leq 2^\circ C$ (e,f); $-1^\circ C \leq T_{\text{RS}} \leq 3^\circ C$ (c,d). Grey shaded areas represent the widest confidence interval among all selections of the parameters.