

Long-term trends (1958–2017) in snow cover duration and depth in the Pyrenees

Juan Ignacio López Moreno, Jean-Michel Soubeyroux, Simon Gascoin, E. Alonso-gonzalez, Nuria Durán-gómez, Matthieu Lafaysse, Matthieu Vernay, Carlo Carmagnola, Samuel Morin

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1 LONG TERM TRENDS (1958–2017) IN SNOW COVER DURATION AND

2 **DEPTH IN THE PYRENEES**

3	Running tittle: Trends in snow cover in the Pyrenees
4 5	López-Moreno, J.I ¹ ; Soubeyroux J.M. ² ; Gascoin, S. ³ ; Alonso-Gonzalez ¹ ; Durán-Gómez, N. ² , Lafaysse, M. ⁴ , Vernay, M. ⁴ , Carmagnola, C. ⁴ ; Morin, S. ⁴
6	1. Pyrenean Institute of Ecology, CSIC. Avda Montañana 1005. Zaragoza, Spain
7	2. Meteo France, Toulouse, France
8 9	3. Centre d'Etudes Spatiales de la Biosphère (CESBIO), Université de Toulouse, CNRS/CNES/IRD/INRA/UPS, 31400 Toulouse, France
10 11	4. Univ. Grenoble Alpes, Université de Toulouse, Météo-France, Grenoble, France, CNRS, CNRM, Centre d'Etudes de la Neige, 38000 Grenoble, France.
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17	Corresponding author:
18	Juan Ignacio López Moreno
19	Pyrenean Institute of Ecology. CSIC
20	Avda Montañana 1005. Zaragoza 50.059. Spain
21	Tel : 0034976369393
22	Email : nlopez@ipe.csic.es
23	

25 Abstract. This study investigated the temporal variability and changes in snow cover duration and the average snow depth from December to April in the Pyrenees at 1500 26 and 2100 m a.s.l. for the period 1958–2017. This is the first such analysis for the entire 27 28 mountain range using SAFRAN-Crocus simulations run for this specific purpose. The SAFRAN-Crocus simulations were evaluated for the period 1980–2016 using 28 in situ 29 30 snow depth data time series, and for the period 2000–2017 using MODIS observations of the snow cover duration. Following confirmation that the simulated snow series 31 32 satisfactorily reproduced the observed evolution of the snowpack, the Mann-Kendall test showed that snow cover duration and average depth decreased during the full study 33 period, but this was only statistically significant at 2100 m a.s.l. The temporal evolution 34 in the snow series indicated marked differences among massifs, elevations, and snow 35 variables. In general, the most western massifs of the French Pyrenees underwent a 36 greater decrease in the snowpack, while in some eastern massifs the snowpack did not 37 38 decrease, and in some cases increased at 1500 m a.s.l. The results suggest that the trends were consistent over time, as they were little affected by the start and end year of the 39 40 study period, except if trends are computed only starting after 1980, when no significant trends were apparent. Most of the observed negative trends were not correlated with 41 42 changes in the atmospheric circulation patterns during the snow season. This suggests that the continuous warming in the Pyrenees since the beginning of the industrial 43 44 period, and particularly the sharp increase since 1955, is a major driver explaining the 45 snow cover decline in the Pyrenees.

46 Keywords: Snow, Trends, Weather Types, Pyrenees

47

48 1. Introduction

49 Seasonal snow is essential for most environmental, hydrological, and geomorphological 50 processes in mountain areas. The insulation effect whereby snow cover protects the 51 underlying substrate from atmospheric temperatures directly controls the thermal 52 regime of soils, and determines the phenological cycles of plants and animals living at 53 high elevations (Kankaanpää *et al.*, 2018; Wang *et al.*, 2018; Serrano *et al.*, 2019). The

snowpack also acts as a natural reservoir of water, formed during the cold season from 54 the accumulated precipitation. The water is released during the melt period, and this 55 governs the timing of river flows and also provides mountain basins with a greater 56 temporal regularity compared with areas dominated by liquid precipitation (Barnett et 57 al. 2005; López-Moreno et al. 2017; Sproles et al. 2018). The state of the snow cover 58 also has clear implications for local communities, as it is a source of major hazards 59 including avalanches, heavy snowfall events, and rain on snow events that affect the 60 security of people and infrastructure (Eckert et al., 2013; Surfleet and Tullos, 2013; 61 62 Navarro-Serrano and López-Moreno, 2017). In addition, its reliable presence is the 63 main factor for various winter tourism activities that have become the primary source of 64 economic income in many mountain regions worldwide (Gilaberte-Búrdalo et al., 2014; Spandre et al., 2019). 65

There is growing concern about the consequences of ongoing climate warming on the 66 mountain snow cover, evident in the growing body of published research devoted to 67 identifying changes in the duration and thickness of snowpack in various regions of the 68 69 world (Hock et al., in press), including the Mediterranean mountains (García-Ruiz et al., 2011; Marchane et al., 2015) and the Alps (Klein et al., 2016; Marty et al., 2017, 70 71 Schöner et al., 2019). Many studies have involved sensitivity analyses and predictions of the impact of various climate scenarios on the snowpack in coming decades (Rasouli 72 73 et al., 2014, 2019; Musselman et al., 2017, Verfaillie et al., 2018, Beniston et al., 2018). However, the study of past snowpack trends is highly constrained by the generally 74 75 limited number of long-term measurements at high elevation sites, where the snowpack plays the most important role (Hock et al., in press). In addition, the stations providing 76 long-term observations are generally unevenly distributed, which makes it difficult to 77 obtain data at regional scales or for entire mountain ranges (Rohrer et al., 2013). 78 79 Another difficulty in snow trend analyses comes from the very high interannual and 80 decadal variability commonly evident in snowpack series (Wegmann et al., 2017); such data can only be adequately analyzed from multidecadal snow series (Mankin and 81 82 Diffenbaugh, 2015). Successful analyses have only been achieved based on data from very dense networks having long-term observations, including that recently published 83 for Switzerland and Austria (Schöner et al., 2019). The European Alps are arguably the 84 European mountain region richest in terms of data availability, and where these type of 85

studies have been conducted (Laternser and Schneebeli 2003; Marty 2008; Morán-86 87 Tejeda et al. 2013; Marke et al., 2018; Schöner et al. 2019; Marty et al., 2017). Studies on the evolution of snow cover based on remote sensing data have provided information 88 on the regional temporal variability of snow cover and snow line elevation, but their 89 limited temporal extent prevents robust assessment of significant trends (Bormann et 90 al., 2018). An alternative to the use of observed snow series is the use of simulated 91 snow series driven by atmospheric reanalyses or observations (Alonso-González et al. 92 2019; van Pelt et al. 2016; Durand et al. 2009; Mote et al., 2018; Marke et al. 2018). 93 94 Although the results can differ from those based on snow observations, they have been shown to adequately reproduce the interannual evolution of the duration and thickness 95 96 of the snowpack (Martin and Etchevers 2005; Durand et al. 2009; Alonso-González et al. 2018). Thus, they can provide long-term and spatially distributed information on the 97 98 snow cover over large areas, enabling regional assessment of snow changes despite 99 limited availability of direct snow cover observational data.

The main objective of this study is to analyze the temporal evolution of snow cover 100 101 duration and average snow depth in the Pyrenees and to assess to which extent the 102 variability is controlled by regional atmospheric circulation. This work also aims to 103 identify spatial differences in the temporal evolution of snow depth series, and to assess the temporal consistency of the trends when different study periods are considered. We 104 105 used simulated snow series from the combined application of the atmospheric reanalysis 106 SAFRAN and the snowpack model Crocus for the period 1958–2017. The temporal consistency of the trends in the snow series was tested by analyzing the dependence of 107 108 the trend patterns on the chosen start and end year of the study time series. We then investigated changes in the frequency of weather types (synoptic situations) over the 109 Pyrenees during each analyzed snow season to assess whether the detected trends were 110 111 dependent on shifts in atmospheric circulation patterns, which are subject to natural variability, or whether the trends could be attributed to other drivers, in particular the 112 113 observed regional climate warming. This is relevant because recent temperature data compiled by weather agencies in France, Spain, and Andorra have shown that the 114 Pyrenees has undergone an increase in annual temperature of 1.3°C since 1955 115 (Pyrenean Observatory of Climate Change, 2019). 116

117

118 **2. Study area**

The Pyrenees is the northern limit of the Iberian Peninsula and creates a natural border 119 between Spain and France. It extends from west to east over 400 km from the Atlantic 120 121 Ocean to the Mediterranean Sea, and is 150 km at its widest point (along a north-south 122 axis) in the central area (Figure 1). It is a transboundary mountain area, including the 123 territory of three countries: France, Spain, and Andorra. The Pyrenean headwaters include the main tributaries of large rivers including the Ebro River, which flows to the 124 Mediterranean Sea, and the Garonne River, which flows to the Atlantic Ocean. The 125 126 Pyrenean range is a key source of runoff meeting the water demands of 7 million people and supplying hydropower and large irrigation areas on both sides of the mountain 127 128 range. The highest peaks of the Pyrenees are in the central and eastern parts of the 129 range, where more than 150 peaks exceeding 3000 m occur (the maximum elevation is 130 reached in the Aneto peak: 3404 m a.s.l.), and large areas are above 2000 m a.s.l.

131 -Figure 1-

132

133 The climate of the Pyrenees is highly variable in the east-west direction, as it reflects the transition from Atlantic to Mediterranean climates. There is also a marked climatic 134 135 difference between the northward and southward slopes along the main axis of the range, with the former being markedly more humid. Annual precipitation ranges from 136 137 600 mm in the dryer areas in the eastern part of the Spanish Pyrenees, to up to 3000 mm in the most humid parts of the western Pyrenees and on slopes facing north and 138 139 northwest in the most elevated central massifs (Serrano-Notivoli et al., 2017). The elevation of the annual 0°C isotherm ranges from 2700 to 3000 m a.s.l. (López-Moreno 140 141 et al. 2019), while the winter 0°C isotherm is located at approximately 1500 m a.s.l., 142 which is the approximate limit for the seasonal snowpack during the coldest months of 143 the year (López-Moreno 2005; Gascoin et al. 2015). Above 2000 m a.s.l. the duration of 144 snowpack often exceeds six months (López-Moreno et al. 2017; Gascoin et al. 2015). 145 Snow is a key resource for the region; López-Moreno and García-Ruiz (2004) estimated that the snowmelt contributes > 40% of the spring runoff for the Ebro headwaters, and 146 147 noted that it is a major complement to dams for synchronizing water demand and water 148 availability, mainly for agriculture and hydropower production. The snowpack has been

also reported to have a significant influence on forest growth and health in several 149 Pyrenean areas at 1300-2300 m a.s.l. (Sanmiguel-Vallelado et al., 2019). There are 150 151 more than 30 ski resorts in the Pyrenees, serving a mean annual population of 5 million 152 skiers, and is the major source of income for many Pyrenean valleys (Gilaberte-Búrdalo 153 et al., 2017). The communities involved depend directly on the quantity and duration of the snowpack, and there is growing concern about its reliability given the climate 154 155 warming projections for the coming decades (Pons et al. 2015), and the extent to which 156 snowmaking and other snow management techniques will be able to address the 157 increasing snow scarcity suggested by climate change scenarios (Spandre et al., 2019).

158

159 **3. Methods**

160 <u>3.1 SAFRAN Crocus simulations</u>

SAFRAN is a reanalysis system combining in-situ meteorological observations, 161 162 remotely sensed cloud cover information and large scale meteorological predictions, in order to provide a consistent and gap-free estimate of the meteorological conditions in 163 164 mountainous regions, here the Pyrenees (Durand et al., 2009). The Pyrenees are decomposed in 23 regions (shown in Figure 1) so called "massifs", considered 165 166 climatologically homogeneous, within which SAFRAN provides meteorological information at hourly time resolution and by elevation steps of 300 m elevation. In 167 168 addition, SAFRAN can also provide meteorological information with adjustments of temperature and radiations considering the exact elevation of a given location and its 169 170 surrounding topographic masks (but the spatial scale of the meteorological analysis 171 remains the massif). In this work, SAFRAN data are used for all massifs in the Pyrenees 172 for 300 m elevations bands on flat terrain. Model simulations used in this study span the time period from 1958 to 2017. This corresponds to a version almost identical to the 173 174 dataset available made available by Vernay et al., (2019). The last corrections made before the publication of this dataset are sufficiently minor to not modify the 175 conclusions of this study. 176

177 The SAFRAN meteorological reanalysis was used to force the detailed snowpack model178 Crocus (Vionnet et al., 2012) which provides estimates of snow cover depth, amount

and various other properties using a multi-layer modelling approach. In this study, we 179 used the spatial configuration for Crocus forced by SAFRAN (massifs/elevations and 180 corresponding to observation stations) used operationally in France in support for 181 avalanche forecasting as described in Morin et al. (2020). The SAFRAN-Crocus 182 183 simulations were also used in the gap filling procedure of 28 long-term snow observation series used in the model evaluation process (section 3.2). For this purpose, 184 at observation stations, observed snow depth data were used to correct the simulated 185 snow depth and re-initialize the following simulations, with the same method as applied 186 187 by Revuelto et al. (2016) with snow depth data from a Terrestrial Laser Scanner. Here, this method was implemented on a daily basis, i.e. simulations were stopped every day, 188 189 snow depth was compared to observed snow depth, and in the case were snow depth 190 values existed, the observed snow depth value was used to adjust the thickness of all 191 simulated snow layers in Crocus so that the snow depth values match, keeping constant the other state variables of each layer (density, temperature, liquid water content, 192 193 microstructure properties). In cases where snow observations were missing, Crocus prognostic variables were left unchanged. In the case where observations indicated 0, 194 195 the Crocus snowpack was removed, and in the case where Crocus predicted 0 snow 196 depth while observations indicated non-zero values, Crocus was reinitialized with a 197 generic, snowpack structure matching the observed snow depth. This typical snowpack 198 structure was generated, using a Crocus model run driven by a SAFRAN forcing file 199 and stopped when the snow depth reached approximately 5 and 50 cm, respectively. Depending on the observed snow depth, either of these two files were used, in order to 200 201 initialize Crocus with a snowpack structure which is consistent with the corresponding 202 snow depth. While this method of direct insertion is not as advanced as proper data 203 assimilation, it allows for gap-filling, taking advantage of the fact that the snow cover 204 evolves with a large auto-correlation in time and that the SAFRAN-Crocus model chain 205 already provides an appropriate estimate of the daily snow cover evolution for trend 206 analysis, based on previous studies (Lafaysse et al., 2013, Durand et al., 2009, Spandre 207 et al., 2019). Meanwhile, this method allowed us to detect and remove outliers in observed snow depth time series, warranting further investigations (combining other 208 data sources, such as temperature, precipitation and wind speed), where the observed 209 snow cover deviates strongly from the simulated one. The outcome of this method is a 210 gap-filled observed snow depth time series, corresponding exactly to the observed 211

values when they are present, and filled with physically consistent values in the case of data gaps, including during the melt season at the end of the winter, when many manual observations are not performed anymore due to the seasonal closure of ski resorts where the observations are carried out by specially trained ski patrollers. Figure 2 shows four examples of simulated snow series with SAFRAN Crocus approach before (Sims) and after applying the re-initialization procedure explained in this section (Sims-Assim).

218

219 -Figure 2-

220 <u>3.2 Evaluation of SAFRAN Crocus simulations</u>

Before doing the trend analysis, the ability of SAFRAN Crocus to reproduce the annual
average snow depth (HS) and annual snow cover duration (SCD) was specifically
evaluated for this study.

224 The simulated HS was compared to the HS obtained from reconstructed snow depth observations at 28 long term stations over the period 01 August 1979 to 31 July 2016 225 226 (hereinafter referred to as "station HS"). These reconstructed snow depth series were considered as "ground truth" since these reconstructed time series do not differ from the 227 228 observations when there are observations, and although the method introduces some 229 artificial errors correlations during the reconstructed periods (after the closure of a ski 230 resort performing observations, the error is constant by construction as the melting rates 231 are identical in the simulation and in the reference). The station HS dataset was made 232 from stations which have a record longer than 20 years and whose average missing data fraction in the daily series is greater than 60% during the months December to April 233 (DJFMA). From these gap-free station series, the mean annual snow depth was 234 235 computed as the mean snow depth during the months DJFMA. The advantage of using 236 this reconstructed station HS instead of the raw HS data is that it enables to compute a 237 mean annual snow depth without having to deal with unevenly distributed data gaps. HS 238 values at a station were compared to HS value in the corresponding massif and 239 elevation band.

240 The simulated annual SCD was evaluated using a daily gap-filled snow cover

climatology from MODIS snow products MOD10A1.005 and MYD10A1.005 (Gascoin 241 et al., 2015). The MODIS dataset covers the period 01 August 2000 to 31 July 2016. 242 For a given hydrological year, the SCD was defined by pixel as the total number of days 243 244 labeled as snow. The same metric was extracted from the SAFRAN-Crocus reanalysis 245 by computing the number of days with a mean daily snow depth higher than 15 cm. We chose 15 cm because it is the optimal threshold identified by Gascoin et al. (2015) from 246 the comparison of MODIS snow products with in situ measurements in the Pyrenees. 247 For a given massif, SAFRAN-Crocus provides one value of the annual snow cover 248 249 duration by elevation band (only the flat slope class was considered). However, MODIS 250 provides one value per 500 m pixel, which means that there is an ensemble of SCD 251 values to compare to this SAFRAN-Crocus SCD value. Hence, the SRTM DEM was 252 used to aggregate all annual SCD values from MODIS pixels by massif and elevation 253 band using the median.

254

255

256 <u>3.3 Statistical analysis</u>

257 From the daily series of snow depth we calculated the annual duration of the snow cover 258 as the number of days from October to September of the following year, and the average snow depth was calculated for the period from December to April, considered 259 typically as the period where a seasonal snowpack can be found in the Pyrenees over 260 261 1500 m a.s.l. (López-Moreno et al., 2005). We selected for further analysis those series corresponding to elevations of 1500 and 2100 m a.s.l. The former was selected because 262 263 it is considered an elevation where the seasonal snowpack is normally present in all the 264 massifs, while at lower elevations the snowpack is ephemeral with multiple melting 265 events during winter (López-Moreno, 2005). Furthermore, the lowest elevation of many 266 ski areas is at approximately 1500 m, which makes it a relevant elevation band on a 267 winter tourism perspective (Pons et al., 2012). At 2100 m a.s.l. the snowpack is much thicker and longerlasting, representing a significant water resource for the Pyrenees. 268 269 Although snow is usually more abundant at higher elevations, the area at elevations of 270 2400 m a.s.l. or higher is relatively small within the Pyrenean range, or absent in the westernmost massifs. The average values for the 23 massifs, for each elevation band 271

considered, were considered as the regional series for the Pyrenees. A varimax rotated principal component analysis (PCA) (Abdi, 2003; Hannachi, 2004) was then performed for each variable (snow cover duration and average snow depth for DJFMA) and elevation (1500 and 2100 m a.s.l.) to facilitate grouping of massifs (those exhibiting the maximum correlation with each PC) having a similar interannual evolution of the snow series.

278 The nonparametric Mann-Kendall (MK) test was used to detect trends in the temporal series (Mann, 1945; Kendall, 1948), using a significance level of p < 0.05. The MK 279 trend test is rank-based, so it can also be applied to series that do not follow a normal 280 distribution or contain outliers (Cunderlik and Ouarda, 2009). The Kendall's tau statistic 281 (range -1 to 1) was used to quantify the strength of the trends. We checked that the 282 283 series were free of lag-1 autocorrelation, and determined that prewhitening of the series was not necessary (Yue and Wang, 2002). To check the temporal coherence of the 284 285 detected trends we calculated a matrix of all possible combinations of starting and ending years (minimum of 20 years length) in the series for each snow variable and the 286 287 corresponding elevation, based on those massifs that showed the highest correlation with the groups that exhibited major trends, as identified by the PCA. This procedure 288 289 enabled us to assess whether the identified trends were dependent on the specific study 290 period considered in our study, or were robust and independent of the interannual 291 variability of the series.

292 <u>3.4 Weather type classification</u>

293 The daily synoptic conditions over the Pyrenees during the study period were calculated 294 using the objective classification method of Jenkinson and Collison (1977), which has been widely applied in climatological studies (see the review of Navarro-Serrano and 295 296 López-Moreno 2017). The method uses a grid of 16 points of atmospheric sea level pressure centered in the central Pyrenees (42°N, 1°E) and spaced at 5° intervals, derived 297 298 from NCEP/NCAR reanalysis (Basnet and Parker, 1997). This enabled calculation of 299 the direction and vorticity of the geostrophic flow, which facilitated classification of each day as cyclonic (C) or anticyclonic (A), and into directional (8) and hybrid (16) 300 301 types. Hybrid types were removed using the approach used proposed by Trigo and 302 DaCamara (2000). The number of occurrences of each weather type recorded each year

from December to April was summed, and the 10 remaining series (C, A, and 8 directions) were reduced using a varimax rotated PCA, as done for the snow series.

305

306 **4. Results**

307 <u>4.1 Evaluation of the simulated snow series</u>

A good correlation (Spearman R = 0.87) was found between the reconstructed mean annual observed snow depth (station HS) and the simulated HS (Fig. 3A); the difference was very small between the two datasets (8 cm). It is important to note that discrepancies between observations and simulations are not only explained by modelling errors but also by the unresolved horizontal intra-massif variability. The performance of the model in relation to snow cover duration obtained from MODIS images was also good (R = 0.96 and a difference of 20.6 days; Fig. 3B).

315

316 -Figure 3-

317

318 <u>4.2 Trend analysis</u>

Figure 4 (A and B) shows the temporal evolution of the regional series for the 23 319 320 massifs in the Pyrenees, calculated as the annual average snow cover duration at 1500 321 and 2100 m a.s.l. and the average snow depth from December to April. The snow cover 322 duration at 1500 m a.s.l. showed large interannual variability, because there were years when the snow did not last more than 1.5 months, while in other years the snow cover 323 324 duration exceeded three months. Over the period 1958-2017 the series showed a 325 negative trend, but this was not statistically significant (p > 0.05). At 2100 m a.s.l. the 326 snow cover duration was also highly variable, with a duration in most years ranging 327 from 4 to 6 months, but the long-term negative trend was statistically significant (p < p0.05). Very similar results were found for the average (DJFMA) snow depth at 1500 328 329 and 2100 m a.s.l. (Fig. 4 C and D). At both elevations the snow depth showed high interannual variability, in most years ranging from 10 to 40 cm at 1500 m a.s.l., and 330

from 60 to 120 cm at 2100 m a.s.l.. Both series showed negative long-term trends, but this was only statistically significant (p < 0.05) for 2100 m a.s.l.

333

334 -Figure 4-

335

336 The results for the regional series shown in Figures 4 did not demonstrate differences in interannual evolution that occurred among the 23 massifs. To assess the variability in 337 snow evolution in the Pyrenees, the PCA classified two main groups of massifs 338 according their similar temporal evolution of snow cover duration at 1500 m a.s.l. (Fig. 339 340 5A); these explained 79% of the variance. Principal component 1 (PC1) was highly 341 correlated with the snow cover duration in the central and western Pyrenees (Fig. 5B). Negative trends clearly dominated in these massifs, especially those located in the 342 westernmost part of the French Pyrenees. Principal component 2 (PC2) reflected the 343 snow cover duration in the eastern massifs, where no trend or a positive trend was 344 apparent. Among the 23 massifs, eight showed statistically significant (p < 0.05) 345 negative trends, and two showed statistically significant positive trends. 346

347 For 2100 m a.s.l. the PCA analysis identified four groups that explained 85% of the 348 observed variance in the interannual evolution of snow cover duration at this elevation 349 (Fig. 5C). PC1 represented the western massifs of the French Pyrenees and the Navarra 350 region on the Spanish side (Fig. 5D). Patterns similar to PC2 2 were found for the 351 eastern massifs; PC3 in central Pyrenees and the PC4 represents some massifs in transition from west to central Pyrenees. Negative Kendall's tau values were found for 352 most of the massifs; 13 of the 23 showed statistically significant negative trends (p < p353 0.05), and none exhibited a statistically significant positive trend. 354

355

356 -Figure 5-

The interannual evolution of average snow depth at 1500 m a.s.l. in the 23 massifs was 358 reflected by three principal components that explained 86% of the observed variance 359 (Fig. 6A). PC1 represented the western and central massifs of the Pyrenees, which had 360 361 generally negative trends, especially in the westernmost French massifs (Fig. 6B); PC2 362 and PC3 represented massifs located to the east. The majority had Kendall's tau values close to 0, but several positive coefficients were found for the easternmost part. Overall, 363 364 five of the 23 massifs showed statistically significant negative trends (p < 0.05), and two showed statistically significant positive trends (p < 0.05). The average snow 365 366 accumulation (DJFM) at 2100 m a.s.l. was also classified into three PCA groups that explained 89% of the observed variance (Fig. 6C). PC1 represented the western French 367 368 massifs, which showed marked negative trends (Fig. 6D), PC2 represented the eastern 369 massifs, and PC3 represented the western Spanish Pyrenees and the central French 370 Pyrenees. Negative Kendall's tau values also dominated the massifs represented by PC2 and PC3, and overall 14 of the 23 massifs showed statistically significant negative 371 372 trends (p < 0.05), and no massif showed a statistically significant positive trend.

373

374 - Figure 6-

375

376 The observed negative trends in snow cover duration and snow depth, found particularly for the western Pyrenees, showed marked temporal consistency and little dependence on 377 378 the study period. The three plots in Figure 7 show that there was a clear dominance of pairs of starting and ending years showing a statistically significant decrease. Only the 379 380 upper right portions of the matrices lack any statistically significant trend, or showed a change in the sign of the Kendall's tau values. Thus, trend analyses starting from 1957 381 to 1980 and finishing in recent years showed statistically significant negative trends that 382 were not affected by decadal fluctuations of the snowpack. However, for series starting 383 384 after 1980 there were no statistically significant trends, and the Kendall's tau coefficients were positive, as can be seen from the temporal evolution of snow depth 385 and duration, shown in the lower right corner of each plot. 386

387

388 -Figure 7-

389

4.3. Atmospheric circulations patterns and their impacts on snow trends

The winter (DJFMA) frequency for the eight weather types classified for the Pyrenees 391 was simplified to four groups of years (PCA; Table 1) that explained 84% of the 392 393 observed variance. PC1 was highly correlated with years characterized by a low 394 frequency of anticyclonic conditions, and a high frequency of advections from the southwest and west. PC2 was highly correlated with years having a high frequency of 395 396 advections from the east and northeast, and a low frequency from the northwest. PC3 was characterized by a low frequency of cyclonic conditions and a high frequency of 397 398 advections from the north. PC4 was characterized by a high frequency of advections from the south and a low frequency from the northwest. 399

400 - Table1-

401

402 Only PC1 showed a statistically significant negative trend (Mann-Kendall tau = -0.53; p < 0.05), which indicated an increase in the frequency of anticyclonic days and a 403 decrease of advections from the northwest and west. The factorial scores for the other 404 three PCs did not shown significant changes (Figure 8), with Mann-Kendall tau values 405 very close to 0. Correlation analysis of PC1 with the groups discriminated according to 406 407 the interannual evolution of snow indices only showed statistically significant correlations with Group 3 (average depth at 2100 m a.s.l.; r = 0.57) and Group 1 (snow 408 409 cover duration at 1500 m a.s.l.; r = 0.29).

410

411 -Figure 8-

412

413 **5. Discussion**

This study showed that the Crocus snow cover model driven by the atmospheric 414 SAFRAN model satisfactorily reproduced the snow evolution over the Pyrenees during 415 the period 1980–2017, enabling the first assessment of long-term trends (1958-2017) 416 417 for this entire mountain range in the absence of comprehensive observational data. This 418 method has previously been used and evaluated in a study in the French Alps (Durand et 419 al., 2009), although it is known that local discrepancies between simulated and observed 420 trends are possible (Lafaysse et al, 2011, Verfaillie et al, 2018). Indeed, simulated 421 trends are affected by various sources of temporal heterogeneity along the simulation 422 period. First, the surface meteorological observations assimilated by the SAFRAN reanalysis have changed over time and it is known that these heterogeneities can 423 424 significantly affect local trends (Vidal et al, 2010). This effect is expected to be reduced when considering larger scale trends, as done in this paper when considering the whole 425 Pyrenees mountain range. Then, the guess of the reanalysis is also heterogeneous over 426 427 time. The two main heterogeneities are due to the onset of satellite observations in the 428 1980s known to be responsible for temporal breaks in ERA-40 reanalysis (Sterl, 2004), 429 and the switch from ERA-40 reanalysis to ARPEGE operational analyses for years after 430 2001. Although these limitations can not be completely removed, it can be reduced in 431 future works by updating the SAFRAN reanalysis by using more recent large scale reanalyses for the guess, such as ERA5. 432

433 Regional series for the Pyrenees revealed declining snow cover duration and average snow accumulation during the study period, but the trends were only statistically 434 435 significant at 2100 m a.s.l., but not at 1500 m. This finding appears counterintuitive, as 436 previous studies have reported greater sensitivity of the snowpack to climate warming 437 in areas near the 0°C isotherm during the snow season compared with colder areas 438 (Pierce and Cayan 2013; Marty et al. 2017). The most likely explanation for our results 439 is the different rates of climate warming reported for the Pyrenees on a seasonal basis. The report recently published by the Pyrenean Observatory of Climate Change (OPCC, 440 2019) and Serrano-Notivoli et al. (2019) notes that warming rates since 1955 have been 441 442 lower during winter months (+0.1 °C per decade; p > 0.05) and greater in spring and fall (+0.3 and +0.2°C, respectively; p < 0.05). A similar seasonal pattern was reported by 443 444 Pérez-Zanón et al. (2017) for the period 1970–2013. No statistically significant change in precipitation occurred during the study period (OPCC, 2019), and its interannual 445

evolution has been found to be closely related to interannual variability in the frequency 446 of weather types (Buisan et al., 2016). In this climate context, snowpacks that lasts after 447 March (the case for the snowpack at 2100 m a.s.l.) are expected to show stronger trends 448 449 than snowpacks restricted to the colder winter months (the case for the snowpack at 450 1500 m a.s.l.). This consideration has not yet been taken into account in interpreting the response of snowpacks to climate warming. The same explanation applies to the evident 451 452 decrease in snow cover duration, affecting to the start and end of the snow cover, and accumulation in the westernmost massifs and massifs to the north of the range 453 454 compared with the remainder of the Pyrenees. Apprndix T1 shows that the snow lasts longer on these massifs, and hence is exposed to the higher warming rates observed 455 456 outside of the coldest months.

457 Some of the significant trends found in this study can also be related to an increase in 458 the occurrence of anticyclonic conditions and a decrease in advections from the west 459 and south west, as evidenced by the statistically significant decrease of PC1. PC1 was 460 clearly linked to the positive evolution of the North Atlantic Oscillation index (NAOi), 461 as previously shown for the Iberian Peninsula by López-Moreno and Vicente-Serrano 462 (2007). In that study it was noted that the interannual evolution of the snowpack in the 463 central Spanish Pyrenees is controlled by the NAO, especially at elevations above 2000 m. Thus, the negative trend in snow accumulation at several observatories in the study 464 465 area can be explained by such changes in atmospheric circulation. Alonso-González et al. (submitted) recently analyzed the effect of the NAO on the snowpack over the entire 466 467 Iberian Peninsula, and showed that it only has an evident effect in this portion of the 468 Pyrenees, and its effect is very limited or nonexistent in the northern massifs and at low elevations throughout the entire range. Thus, many of the negative trends reported in 469 470 this study cannot be directly explained by changes in atmospheric circulation patterns, as the other patterns identified did not show any significant trend. In the absence of 471 472 clear trends for precipitation, this finding strongly suggests that the main driver of the 473 decline in the snowpack in the Pyrenees is long-term climate warming.

The major trends in the Pyrenees during the study period seemed to be consistent over time and not significantly affected by the selected start and end years when long-term time series were considered. However, for analysis periods starting after 1980, no significant trends in the snowpack were found. A very similar finding has been reported for the Swiss and Austrian Alps, where only small changes since the end of the 1980s have been observed (Marty, 2008; Schöner *et al.*, 2019). This result illustrates the importance of using sufficiently long observation records when assessing trends of snow series (Fassnacht and Hulstrand, 2015).

For the Pyrenees, the temporal series of temperature evolution in the evolution provides 482 483 an explanation (OPCC, 2019). The temperature increase since 1955 is mostly evident in the stark contrast between the cold decades of the 1960s and 1970s, and the much 484 warmer following decades. However, the temperature trends after 1980 have been 485 weaker or nonexistent (López-Moreno et al., 2019). This explains why some studies 486 using snow data for the Spanish Pyrenees since 1986 have found no trends in the 487 snowpack (Buisan et al. 2015; Morán-Tejeda et al. 2017); for this specific part of the 488 489 Pyrenees this finding is reinforced by the relatively high frequency of winters having 490 negative NAOi values (Añel et al. 2014; Buisan et al. 2016), which is associated with 491 frequent humid conditions in the massifs located southwards.

492

493 **6.** Conclusions

494 The main findings of this work can be summarized as follows:

1. SAFRAN-Crocus satisfactorily reproduced the main space and time variations in the
snow cover in the Pyrenees from 1980 to 2017 based on in-situ and remote sensing
observations, providing confidence in the trends in simulated values from 1958 to 2017
although the temporal homogeneity of the data is not fully guaranteed.

2. In general, the snowpack decreased during the period 1958–2017, and the decline has
been clearer at 2100 m a.s.l. This is consistent with the climate trends detected in the
study, as warming during the study period was greater in fall and spring compared with
winter (when the snowpack is limited at 1500 m).

503 3. There was marked spatial variability among massifs, and this differed for the two 504 analyzed indices (snow cover duration and average snow depth DJFMA) and the two 505 analyzed elevation bands (1500 and 2100 m a.s.l.). Overall, the western Pyrenees 506 showed the most statistically significant decline in the snowpack. 4. Most of the significant trends detected showed high temporal consistency during the
study period, and low dependence on the period selected, provided this was sufficiently
long. The results suggest that only trends for snow accumulation at 2100 m a.s.l. in the
central Pyrenees are linked to changes/variability in atmospheric circulation patterns
(increase in the NAOi).

5. This study highlights the need for using long-term records and evenly distributedspatial data in the analysis of snow trends.

514

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Appendix T1. Average snow duration and snow depth (DJFMA) for each analyzed massif, their corresponding group and Tau-Kendall values for the period 1958-2017. Bold numbers indicate statistically significant trends.

	Average values				Kendall-Tau and classified group							
Massif	Dur. 1500	Dur. 2100	Depth 1500	Depth 2100	Dur. 1500	PC	Dur. 2100	PC	Depth 1500	PC	Depth 2100	PC
Pays	106	150	0.31	1.70	-0.42	1	-0.10	1	-0.38	1	-0.33	1
Basque Aspe-Ossau	134	150	0.37	1.71	-0.27	1	-0.17	1	-0.22	1	-0.23	1
Hte. Bigorre	122	148	0.18	0.93	-0.15	1	-0.18	1	-0.10	1	-0.20	1
Aure- Louron	112	148	0.19	0.93	-0.21	1	-0.15	1	-0.19	1	-0.28	1
Luchonnais	116	148	0.23	1.03	-0.22	1	-0.17	1	-0.18	1	-0.26	1
Couserans	126	148	0.25	1.11	-0.19	1	-0.17	3	-0.08	1	-0.11	3
Aran	99	146	0.17	0.84	-0.09	1	-0.16	3	0.07	2	-0.17	1
Hte. Ariege	122	146	0.20	0.91	-0.15	2	-0.21	2	-0.14	2	-0.18	3
Andorre	66	126	0.04	0.31	0.01	1	-0.23	3	0.06	1	-0.25	3
Orlu- StBarthe	118	146	0.19	0.94	-0.10	2	-0.23	2	-0.09	2	-0.21	2
Capcir- Puymorens	67	127	0.06	0.36	-0.02	2	-0.18	2	0.00	3	-0.18	2
Cerdagne	85	120	0.05	0.40	-0.18	2	-0.13	4	0.17	3	-0.01	2
Navarra	108	149	0.31	1.72	-0.31	1	-0.21	1	-0.18	1	-0.19	3
Jacetanía	109	149	0.22	1.44	-0.17	1	-0.13	4	-0.11	1	-0.20	3
Gállego	120	148	0.24	1.40	-0.10	1	-0.08	4	0.00	1	-0.15	3
Sobrarbe	127	147	0.18	1.09	-0.06	1	-0.02	4	0.06	1	0.03	3
Ésera	96	147	0.17	0.94	-0.11	1	-0.18	3	-0.03	1	-0.14	3
Ribagorza	85	146	0.16	0.89	-0.11	2	-0.14	3	0.01	2	-0.17	2
Pallaresa	100	139	0.09	0.50	-0.13	2	-0.11	2	0.17	2	-0.02	2
Perafita	60	125	0.05	0.35	-0.01	2	-0.31	2	0.02	3	-0.25	2
Teruel- Frese	48	123	0.06	0.40	0.18	2	-0.06	2	0.17	2	0.09	2
Cadi- Moixero	52	126	0.05	0.39	0.17	2	-0.22	2	0.11	3	-0.11	2
Prepirene	75	130	0.07	0.48	0.08	2	-0.21	4	0.04	3	-0.10	3

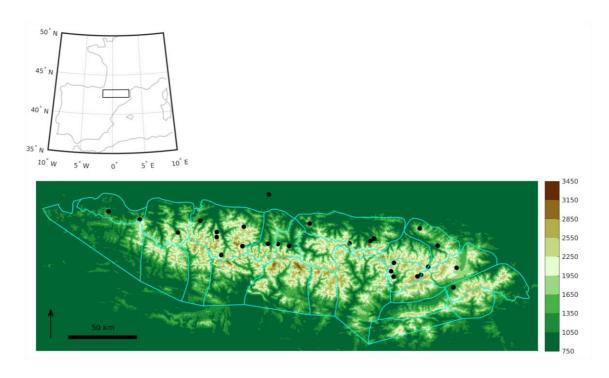
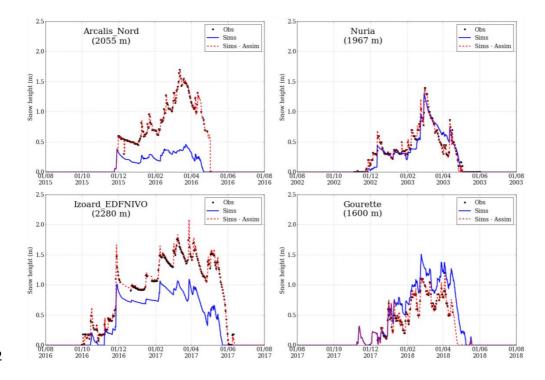




Figure 1. The study area, and delimitation of the 23 massifs into which the Pyrenees
was divided for the SAFRAN-Crocus simulation. Black dots indicate the location of the
snow depth observation stations (full list given in the Appendix T1).



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773 Figure 2. Examples of observed and simulated snow series with SAFRAN Crocus

approach in four observatories before (Sims) and after applying the correction

procedure explained in this section (Sims- Assim).

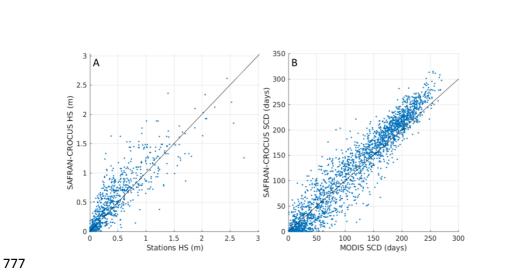


Figure 3. A: Comparison of the mean annual snow depth, as simulated by SAFRANCrocus and from the reconstructed station dataset. B: Comparison of the mean annual
snow cover duration, as simulated by SAFRAN-Crocus and from the MODIS dataset.
Each dot corresponds to a triplet (massif, elevation band, hydrological year).

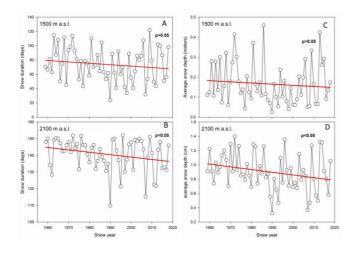


Figure 4. Evolution of annual snow cover duration (A and B) and average snow depth

from December to April (C and D) at 1500 and 2100 m a.s.l. respectively.

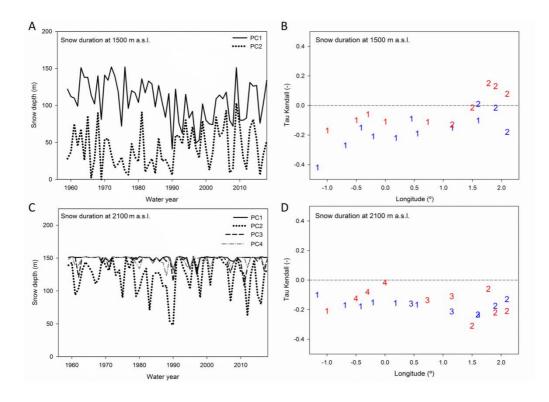
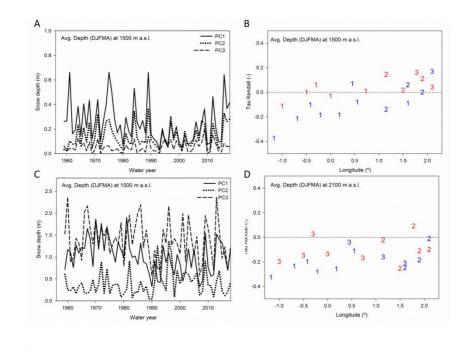


Figure 5. Temporal evolution of the factorial scores for the groups of massifs identified
by PCA for the annual evolution of snow cover duration at 1500 m a.s.l. (A) and 2100
m a.s.l. (C). Kendall's tau coefficients for the massifs classified in the various groups
(numbers) from west to east at 1500 m a.s.l. (B) and 2100 m a.s.l. (D). Those massifs
located north of the main divide are shown in blue, and those to the south are shown in
red.



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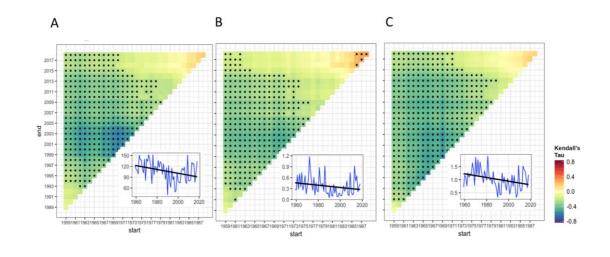
Figure 6. Factorial scores for the groups of massifs identified by principal component

analysis for the evolution of snow depth (DJFM) at 1500 m a.s.l. (A) and 2100 m a.s.l.

798 (C). Kendall's tau coefficients for the massifs classified in the various groups (numbers)

from west to east at 1500 m a.s.l. (B) and 2100 m a.s.l. (D). Those massifs located north

800 of the main divide are shown in blue, and those to the south are shown in red.



803 Figure 7. Results of the Mann-Kendall test applied to the observatories that exhibited the largest correlation with those principal components characterized by the strongest 804 805 negative trends. The matrix shows all possible combinations of start and ending dates 806 involving at least 20 years duration during the period 1958-2017. A) Aspe-Ossau massif for PC1 for snow cover duration at 1500 m a.s.l. B) Navarra massif for PC1 for 807 808 average snow depth (DJFMA) at 1500 m a.s.l. C) Luchonnais massif for PC1 for average snow depth (DJFMA) at 2100 m a.s.l. Colors indicate the magnitude of the 809 810 Kendall's tau, and points mark the series that showed statistically significant trends (p < 811 0.05).

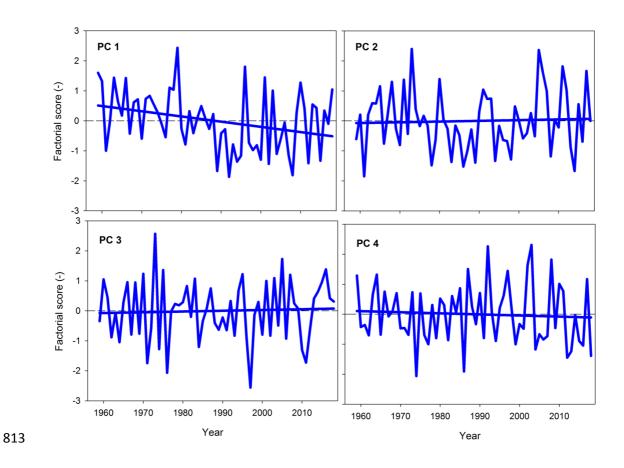


Figure 8. Temporal evolution of the factorial scores for the four principal components
that summarized the frequency of weather types over the Pyrenees during the study
period.