



Spatial variability in the seasonal precipitation lapse rates in complex topographical regions – application in France

Valentin Dura^{1,2}, Guillaume Evin², Anne-Catherine Favre², and David Penot¹

¹EDF-DTG, 38950 Saint-Martin-le-Vinoux, France

²Univ. Grenoble Alpes, INRAE, CNRS, IRD, Grenoble INP, IGE, 38000 Grenoble, France

Correspondence: Valentin Dura (valentin.dura@edf.fr)

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Abstract. Seasonal precipitation estimation in ungauged mountainous areas is essential for understanding and modeling a physical variable of interest in many environmental applications (hydrology, ecology, and cryospheric studies). Precipitation lapse rates (PLRs), defined as the increasing or decreasing rate of precipitation amounts with the elevation, play a decisive role in high-altitude precipitation estimation. However, the documentation of PLR in mountainous regions remains weak even though their utilization in environmental applications is frequent. This article intends to assess the spatial variability and the spatial-scale dependence of seasonal PLRs in a varied and complex topographical region. At the regional scale (10 000 km²), seven different precipitation products are compared in their ability to reproduce the altitude dependence of the annual/seasonal precipitation of 1836 stations located in France. The convection-permitting regional climate model (CP-RCM) AROME is the best in this regard, despite severe precipitation overestimation in high altitudes. The fine resolution of AROME allows for a precise assessment of the influence of altitude on winter and summer precipitation on 23 massifs at the sub-regional scale (~1000 km²) and 2748 small catchments (~100 km²) through linear regressions. With AROME, PLRs are often higher in winter at the catchment scale. The variability in the PLR is higher in high-altitude regions such as the French Alps, with higher PLRs at the border than inside the massifs. This study emphasizes the interest of conducting a PLR investigation at a fine scale to reduce spatial heterogeneity in the seasonal precipitation–altitude relationships.

1 Introduction

Precipitation lapse rates (PLRs) refer to the increasing or decreasing rate of precipitation amounts with elevation. The rise in the moist air above an orographic barrier such as hills or mountains generates precipitation that affects the ice cover and the precipitation amount contained in snowpacks (Bales et al., 2006; Viviroli et al., 2007; Mott et al., 2014; Dozier et al., 2016; Wrzesien et al., 2019). Therefore, the PLR is the main factor that controls the water budget of high-altitude catchments (Jiang, 2003) and partly explains the freshwater supply in the summer season. On shorter timescales like hourly or daily scales, PLRs help us understand the physical mechanisms of orographic precipitation responsible for natural hazards such as avalanches, landslides, and floods (Caracena et al., 1979; Caine, 1980; Conway and Raymond, 1993; Buzzi et al., 1998; Panziera et al., 2015). PLRs can also be of importance for hydrological modeling for which precipitation inputs are often spatialized with geostatistics models (Daly et al., 1994; Gottardi, 2009; Frei and Isotta, 2019), or they can be distributed by altitudinal bands using assumed PLR values (Bergström, 1992; Ragettli and Pellicciotti, 2012; Markstrom et al., 2015; Garavaglia et al., 2017; Ruelland, 2020; Kumar et al., 2022) to complete the hydrological balance (Oudin et al., 2006) in high-altitude catchments. The investigation of PLR is a topical and crucial issue in multiple domains such as energy production, agriculture, tourism, the condition of ecosystems, and risk management (Pimentel et al., 1997; Gössling et al., 2012).

The relationship between altitude and precipitation accumulation is not straightforward on small timescales and depends on the precipitation type (stratiform vs. convec-

tive), wind, and flow directions (Sevruk and Miegliitz, 2002; Schättli, 2013). PLR at the hourly scale may even exhibit a negative trend during extreme events (Formetta et al., 2022) due to the drying of the air mass. When aggregated over extended periods, the relationship between precipitation and elevation becomes simpler to model, generally with a precipitation increase as elevation increases (Barrows, 1933; Spreen, 1947; Schermerhorn, 1967; Smith, 1979). However, PLR values can differ among seasons and regions due to different weather processes (Napoli et al., 2019; Ménégoz et al., 2020). In the Alps, large-scale processes mainly generate winter precipitation. Convective processes are more frequent in summer.

In a hydrology framework, PLRs refer to the slope of the linear regression between precipitation and elevation on spatial areas (e.g., Sevruk, 1997; Gottardi, 2009; Ogrin and Kozamernik, 2020; Avanzi et al., 2021; Bell et al., 2022) rather than on vertical atmospheric columns. In some cases, the relationship in high-altitude regions appears to be non-linear with a threshold impact (Schättli, 2013; Napoli et al., 2019) due to the drying of the air masses, and the use of quadratic regression is then recommended (Mahmood et al., 2021).

PLRs are often estimated from ground measurements from rain gauges, observations of snow water equivalent (SWE) (Avanzi et al., 2021) derived from manual coring or radiation, and snow/rain gauge totalizer (Gottardi, 2009) measures. Using only ground stations to estimate PLRs requires a sufficient coverage of high-elevation areas, which is rarely met in practice (Hofstra et al., 2010). High-elevation stations are prone to precipitation undercatch (Groisman and Legates, 1994; Pollock et al., 2018) and snow redistribution. Strong winds coupled with a significant amount of solid precipitation might even induce an annual underestimation of up to 25 % (Sevruk, 1997). Moreover, the robustness of the regressions and the local influences of altitude on precipitation must be balanced because of the irregular spatial sampling of punctual observations (Gottardi, 2009). In practice, in sparse rain gauge regions, the PLR can only be computed on vast areas and fails to reflect the local relationship between altitude and precipitation.

As an alternative, gridded precipitation products can help analyze the spatial variability in the PLR at a fine spatial scale. Radar products that indirectly measure precipitation from reflectivity have been extensively used in hydrology (e.g., Ochoa-Rodriguez et al., 2019). However, radar data are inherently biased in mountainous regions due to ground echoes and beam blockage (e.g., Berne and Krajewski, 2013), which results in an underestimation of PLR (Faure et al., 2019). Satellite precipitation products, despite certain limitations, are another easily accessible source of data. The biases in mountain regions and the relatively low resolution of the data present challenges in calculating PLR from satellite data (Li et al., 2017). Composite products merging radar or satellite to rain gauge data (Champeaux et

al., 2009; Nie et al., 2015; Nguyen et al., 2020) can also be an option for the assessment of PLRs. However, the latter products are sensible on the density of the rain gauge network and suffer where no rain gauges are available (Silverman et al., 2013; Shen et al., 2018; Frei and Isotta, 2019). Regional climate models (RCMs) have been used in numerous studies to evaluate the variability in the PLR at regional scales (e.g., Kotlarski et al., 2012; Cuo and Zhang, 2017; Ménégoz et al., 2020). All results show that PLRs obtained with RCM data are larger and more spatially varying than those using rain gauges. Convection-permitting regional climate models (CP-RCMs) have recently emerged as an appealing tool for producing fine-resolution (1–4 km²) climate simulations (Rockel et al., 2008; Brousseau et al., 2016; Keuler et al., 2016; Belušić et al., 2020). CP-RCMs are a refined extension of RCM where deep-convection parameterization, a dominant source of precipitation error (Hohenegger et al., 2008; Foley, 2010; Kendon et al., 2012), is turned off. To produce a long series of CP-RCM simulations, including precipitation, a numerical weather prediction (NWP) model is often driven by a RCM or a satellite-based reanalysis. CP-RCMs produce better precipitation intensities and frequencies than RCMs during heavy events (Ban et al., 2021; Cailaud et al., 2021). CP-RCMs represent orographic precipitation more accurately than RCMs because they explicitly resolve deep convective processes and include a finer representation of the topography (Lucas-Picher et al., 2021). These improvements occur especially in summer when convective events play a major role (Ban et al., 2021). Today, CP-RCMs are considered more reliable than reanalysis (Lundquist et al., 2019) in ungauged mountainous areas, despite the current precipitation overestimation (Gerber et al., 2018; Dallan et al., 2023).

This paper explores the spatial variability in the winter and summer PLR across a complex topographical region with a dense network of 1836 stations used as ground truth. The first objective is to compare seven products from different categories (interpolator, reanalysis, satellite, radar, and CP-RCM) with respect to their ability to capture the relationship between observed annual/seasonal precipitation and altitude. The second goal is the investigation of the spatial variability in the winter and summer PLR values at the sub-regional (≈ 1000 km²) and catchment (≈ 100 km²) scales using a gridded precipitation product without ground station assimilation. Nesting catchments in the sub-regions allows the study of the spatial-scale dependence of PLR to provide guidelines for future precipitation interpolation studies. To our knowledge, only Jiang et al. (2022) explore these questions in the Third Pole. In a sparse rain gauge region, the authors found an accurate PLR reproduction with a down-scaled ERA5 precipitation product (Jiang et al., 2021). They highlighted the spatial variability in the PLR at the catchment scale and emphasized the importance of conducting PLR studies at the finest possible spatial scale.

This paper is organized as follows. Section 2 describes the study domain and the available precipitation products. Section 3 presents the method used to derive PLR. Section 4 shows the results of the reproduction and spatial variability in the PLR by the different products considered. Section 5 compares the findings to the literature and discusses some perspectives. Section 6 gives the conclusions.

2 Domain and data under study

2.1 A complex mountainous region

Figure 1a shows the study domain which corresponds to the intersection of the spatial extents of the seven gridded precipitation products. The topography of the region is complex and varied, including three major mountain chains: the Alps, the Pyrenees, and the Massif Central. The Massif Central is a mid-mountain range reaching 1885 m in the center of the southern half of France and covers 85 000 km², which makes it the largest mountainous region of the country (15 % of the total surface area of France). The Massif Central is bordered by the Rhône Valley to the west and the Mediterranean coastal plain to the south. This region is mainly composed of large plateaus ranging in altitude from 600 to 900 m and enclosed by mountains with circular summits corresponding to extinct volcanoes. The altitude presents an asymmetrical profile with summits in the south and the east (Cévennes) and less elevated areas in the northwest. The landscapes are varied and contain limestone plateaus cut by deep canyons, mountain peaks, and deep river valleys. The Massif Central is exposed to the west wind and captures the precipitation coming from the Atlantic Ocean.

The Alps are the highest mountain range in Europe, reaching 4808 m at Mont Blanc. The chain has a 40 000 km² surface area and is 1200 km long, crossing seven countries, namely France, Switzerland, Italy, Germany, Liechtenstein, Austria, and Slovenia. The French Alps are separated from the Massif Central by hills and are divided into the northern and the southern parts. The Northern Alps are higher and have large valleys. In the Southern Alps, valleys are narrower. The climate of the French Alps has three main influences. The moist air can come from the west Atlantic. The warm air comes from the Mediterranean and blows northward. In winter, easterly weather fronts coming from the Italian Alps can bring large amounts of precipitation to the east of the French Alps (Garavaglia et al., 2010). Generally, these meteorological events do not reach the foothills of the French Alps. The Northern Alps are more rainy than the Southern Alps, which have almost 300 d yr⁻¹ of sunshine. The French Alps is a territory of high-altitude and complex topography. The significant variations in mountain elevation and exposure result in the coexistence of multiple climates at small spatial scales. Mountains create micro-climatic conditions and influence daily weather. The Jura Mountains are a sub-alpine

mountain range located in the north of the French Alps. They cover a surface area of 5000 km² and reach 1700 m. The Jura Mountains are a mid-mountain range composed of plateaus with a mean altitude of 600 m.

The Pyrenees form a natural border between France and Spain by joining the Atlantic Ocean to the Mediterranean Sea over 430 km. It covers 19 000 km² and reaches 3200 m on the French side, making it the second-highest mountain range in France. Most valleys are running north–south. The Spanish part is more arid and rocky than the French part, which catches the precipitation from the Atlantic. The French Pyrenees splits into the Atlantic, central, and eastern parts. The relief rises steeply from sea level to hills in the Atlantic and eastern parts. The central Pyrenees hosts sharp peaks but is more rounded than the Alps because of erosion. The Atlantic part gets more precipitation because of rainy north-west winds. The wind from the southwest brings warmer air in summer. The topography is diverse, with snow-capped mountains, deep green valleys, and coastal hills. The remainder of France consists of plains and is not included in this study.

In this study, we consider three spatial scales: the regional scale (10 000 km²), the sub-regional scale (1000 km²), and the catchment scale (100 km²). The sub-regional scale consists of areas named “massif”, which form a set of continuous reliefs often separated by rivers and valleys. An ensemble of massifs constitutes a mountain range (Alps and Pyrenees). In the French Alps, 23 massifs have been identified through climatological homogeneity of precipitation in Pahaüt (1991). PLR computation will be conducted at the massif (climatological homogeneity) and catchment (hydrologic interpretation) scales. The surface areas of the massifs range from 450 to 1600 km², with a median value of 870 km². The 2748 catchments collected from Sandre (2020) have a mean surface area ranging from 0.009 to 1000 km², with a median value of 100 km². Figure 1c illustrates the massifs and the catchments analyzed in this study. The altitude is represented with a digital terrain model of 1 km resolution. The study domain is divided into squares of 1 km² that are referred to as “pixels”.

2.2 A dense network of stations

Figure 1b presents the dense rain gauge network of 1836 stations used in this study, with an average density of one rain gauge per 160 km². Electricité de France (EDF, the main electricity provider in France) and Météo-France provide 673 and 1163 stations, respectively. We aggregate available daily precipitation data to obtain seasonal and annual precipitation. Figure 1d shows the number of rain gauges per altitudinal band. A small number of rain gauges is available above 2000 m, causing an underrepresentation of high-altitude regions. To overcome this limitation, EDF has implemented a snow rain gauge totalizer (Nivo Pluviomètre Totalisateur, NPT, in French) device (Gottardi, 2009). NPTs collect an-

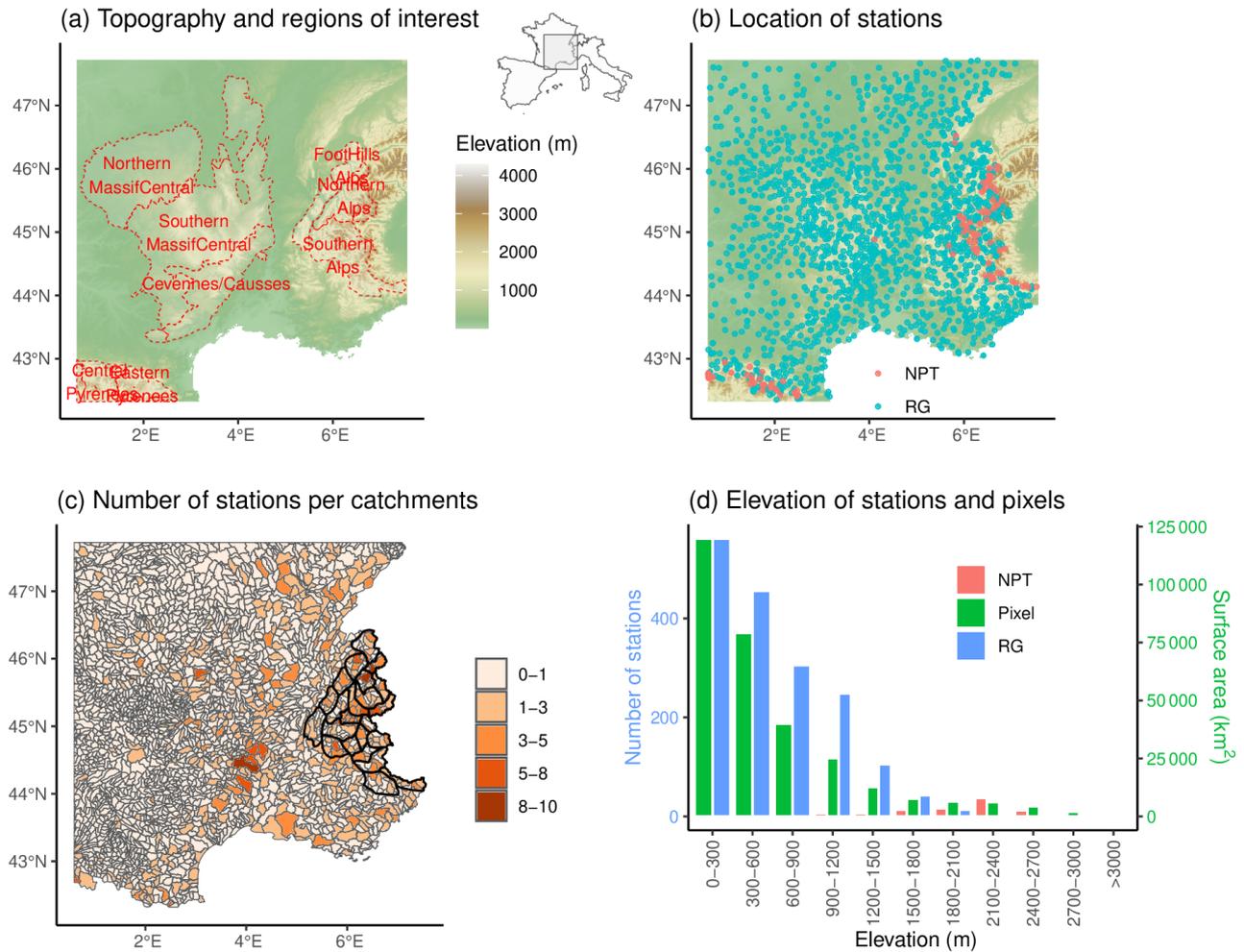


Figure 1. (a) Topography of France and boundaries of eight sub-regions of interest. The extent of the study domain is located in France and is bordered by Spain, Italy, and Switzerland. (b) Location of stations; blue points refer to rain gauges (RGs) and red points to NPTs (see Sect. 2.2). (c) Catchments and the corresponding number of stations. Boundaries of massifs located in the Alps are also indicated in black. (d) Histogram of the altitude of stations (rain gauges (RGs), and NPTs) and of the covered surface area (in green).

nual precipitation in a small orifice protected from the wind by a large collar. Snow falling in the orifice feeds a reservoir containing a saline solution able to melt solid precipitation. The operating period ranges from 1945 to 1977 and differs among NPTs. We restricted the period to 1957–1973, during which the spatial density of NPT is at its maximum. An exploratory analysis of the annual precipitation from NPTs removes outliers and missing values. Following this analysis, we keep 130 NPTs among 216 for the study. NPTs are primarily found in the Mont Blanc region and are more uncommon, for instance, in the center of the Pyrenees. Figure 1c illustrates the irregular number of weather stations per catchment. Mountainous regions like the Alps and the Cévennes are more covered than plains, where weather processes are less subject to spatial variability. Nevertheless, the Southern Alps, the Massif Central, and the Pyrenees have a complex topography and few stations. The sparse stations

in some mountainous regions and their location in valleys largely constrain the ability to compute PLR using only precipitation aggregated from rain gauges.

In France, precipitation patterns differ among seasons. Thus, in this study, four seasons are considered: winter (December, January, February – DJF), spring (March, April, May – MAM), summer (June, July, August – JJA), and autumn (September, October, November – SON).

2.3 Gridded precipitation products

This study explores the ability of multiple gridded precipitation products to assess the relationship between annual precipitation and altitude. Precipitation from gridded products are available at daily, hourly, or sub-hourly timescales and aggregated to obtain seasonal precipitation. These products are briefly described hereafter and are summarized in Table 1.

Table 1. Summary of the characteristics of the gridded precipitation products used in this study, namely the available period, domain, horizontal resolution, frequency, type of observations, provider, and reference.

	Available period	Domain	Horizontal resolution	Frequency	Type of observations	Provider	Reference
ERA5-Land	1950–2023	World	9 km	Hourly	Reanalysis ECMWF replay	ECMWF	Muñoz-Sabater et al. (2021)
CERRA-Land	1984–2023	Europe	5.5 km	Daily	Reanalysis	C3S	Le Moigne (2021)
PDIR	2000–2023	60° N–60° S	4 km	Hourly	Reanalysis satellite	CHRS	Nguyen et al. (2020)
SERVAL	2006–2020	France	1 km	5 min	Radar	Météo-France	Champeaux et al. (2009)
COMEPHORE	2007–2020	France	1 km	Hourly	Reanalysis rain gauges + radar	Météo-France	Champeaux et al. (2009)
SPAZM	1950–2020	France	1 km	Daily	Reanalysis rain gauges	EDF	Gottardi (2009)
AROME	1982–2018	Pan-Alps	2.5 km	Hourly	CP-RCM	Météo-France	Caillaud et al. (2021)

2.3.1 ERA5-Land

ERA5-Land (Muñoz-Sabater et al., 2021) is a global reanalysis that operates by simulating atmospheric and surface variables, including hourly precipitation. It uses the ERA5 atmospheric fields downscaled with linear interpolation at the resolution of 9 km. ERA5-Land assimilates satellite data and does not rely on rain gauges. The quality of ERA5-Land precipitation is dependent on the density of satellite and radar networks used in the assimilation process (Hassler and Lauer, 2021). As a result, the quality is better in Central Europe and the USA than in tropical oceans. The improved horizontal resolution of 9 km compared to 31 km (ERA5) or 80 km (ERA-Interim) is a benefit of ERA5-Land (Muñoz-Sabater et al., 2021). This results in a better representation of spatial precipitation patterns (Gomis-Cebolla et al., 2023).

2.3.2 CERRA-Land

Copernicus European Regional ReAnalysis Land (CERRA-Land) (Le Moigne, 2021) is the most up-to-date reanalysis available in Europe since September 1984 at a horizontal resolution of 5.5 km. It uses ERA5 as lateral boundary conditions to run the atmospheric model HIRLAM ALADIN Regional/Mesoscale Operational NWP in Euromed (HARMONIE) every 3 h at a native resolution of 5.5 km. The analysis is given to the MESCAN system (Soci et al., 2016) to produce daily precipitation fields by merging it with nearly 8000 daily rain gauge measurements through optimal interpolation. Le Moigne (2021) found that CERRA-Land leads to a better representation of the seasonality of snow depth in the Alps compared to ERA5-Land.

2.3.3 PDIR-NOW

PERSIANN Dynamic Infrared–Rain Rate (PDIR-NOW; Nguyen et al., 2020) is a satellite-based product using high-frequency (15 min) sampled infrared imagery at a 4 km spatial resolution. PDIR-NOW estimates precipitation based on empirical cloud-top temperature–precipitation rate relationships. Errors resulting from this method are corrected by calibrating the empirical relationships regionally based on monthly precipitation climatology. PDIR-NOW has better di-

urnal cycle representation, rain/no rain day estimation, and regional precipitation patterns compared to the other PERSIANN family products (Nguyen et al., 2020). The inter-annual, annual, and seasonal precipitation is less biased with PDIR-NOW (Huang et al., 2021; Uysal, 2022). In the following, PDIR-NOW will be referred to as PDIR.

2.3.4 SERVAL

The *Système d'élaboration des produits radar et de visualisation centralisée* (SERVAL, previously PANTHERE; Champeaux et al., 2009) product provides precipitation amounts at a frequency of 5 min and a spatial resolution of 1 km by merging individual radar fields. Before any radars are combined, each one is individually calibrated using data from hourly rain gauges with a mean hourly correction ratio applied uniformly across the whole radar coverage. If a pixel falls under the coverage of multiple individual radars, then the selected precipitation corresponds to the maximum of precipitation seen by the considered radars. As already mentioned, radars are affected by many sources of potential errors, such as beam blocking and ground echoes. Ground echoes are removed using pixel precipitation probabilities, which are derived from a cloud classification filter. Bright band and signal attenuation phenomena are also corrected.

2.3.5 COMEPHORE

The *COMbinaison en vue de la Meilleure Estimation de la Précipitation HORaire* (COMEPHORE; Champeaux et al., 2009) product is a precipitation reanalysis based on the combination of the previous SERVAL product with a dense rain gauge data set of more than 4000 observations. Kriging of the rain gauge values is applied to obtain the stratiform part, while the convective part involves the radar precipitation recalibration with the rain gauge amounts. The main idea underlying this product is to combine the coherent spatial structure of the radars with the accuracy of rain gauge measurements to provide an hourly precipitation field at 1 km resolution. COMEPHORE underestimates mountain annual precipitation (Rouzeau, 2013; Roger, 2017) since the altitude is not considered at any modeling stage.

2.3.6 SPAZM

The SPATialisation en Zones de Montagne (SPAZM; Gottardi, 2009) product is a precipitation interpolator that integrates rain gauge values with a meteorological guess conditioned by topography and the weather type of the day. Eight weather situations are identified based on geopotential fields (Garavaglia et al., 2010), and eight corresponding mean daily precipitation fields are created through local linear regression between the mean daily observed precipitation at stations and the altitude at the daily timescale and 1 km resolution. For a given day, the corresponding mean daily precipitation field is modified using the precipitation of the current day. SPAZM incorporates 2201 various stations from EDF (rain gauges and NPTs), Météo-France, MeteoSwiss, Arpa Piemonte, and Instituto Nacional de Meteorología (INM) networks. SPAZM used fewer but higher stations than COMEPHORE. Its ability to provide accurate estimates of yearly precipitation makes it a valuable tool for hydrological applications (Gottardi, 2009; Ménégoz et al., 2020; Ruelland, 2020).

2.3.7 CNRM-AROME

The CNRM-AROME model is a CP-RCM based on the non-hydrostatic, convective-scale, limited-area model called Applications de la Recherche à l'Opérationnel à Mésio-Echelle (AROME) and has been used for the national weather prediction by Météo-France since 2008 (Seity et al., 2011; Brousseau et al., 2016). There are 37 years of CNRM-AROME simulations available at a horizontal resolution of 2.5 km and at the hourly timescale (Caillaud et al., 2021) for an Alpine domain defined in the flagship pilot study of the Coordinated Regional Climate Downscaling Experiment (CORDEX-FPS; Coppola et al., 2020). The Atlantic Pyrenees are not contained in the domain. The version of AROME applied to obtain this long simulation is related to cycle 41t1 (Termonia et al., 2018) and corresponds to the version in operational use at Météo-France between December 2015 and December 2017. In the remainder of the paper, the CNRM-AROME CP-RCM will be referred to as AROME. The initial atmospheric conditions of AROME are provided every hour by a CNRM-ALADIN RCM simulation, itself driven by the ERA-Interim reanalysis. The fine resolution of the model enables a good representation of the deep convection scheme despite the absence of data assimilation. Precipitation from this version has been evaluated recently in multiple studies (Ban et al., 2021; Caillaud et al., 2021; Monteiro et al., 2022). Heavy precipitation is better represented than in the RCM CNRM-ALADIN but suspicious snow accumulation is pointed out in high altitudes. A first pre-processing stage is required to address large errors in precipitation seasonal amounts at the pixel scale. An iterative technique is applied to deal with the problematic pixels, whereby the abnormal values are replaced with the weighted (according to the altitude) average of the precipitation amount of the nearby

pixels. A pixel is qualified as problematic if a difference of more than 500 mm in seasonal precipitation is noticed compared to a neighboring pixel.

Some gridded precipitation products present non-homogeneous data because of their large temporal depths. Most radars have been integrated since 2006, and others have been gradually incorporated since 2015 in mountainous regions (Beck and Bousquet, 2013). The use of SERVAT and COMEPHORE is therefore tainted with temporal non-homogeneity. Station density also affects the temporal homogeneity of COMEPHORE, CERRA-Land, and SPAZM. Changes in instrumentation have been corrected in COMEPHORE and SPAZM with the homogenization of rain gauge precipitation (Gottardi, 2009; Mestre et al., 2013).

Figure 2 displays the mean annual precipitation for all gridded precipitation products. Figure S1 in the Supplement provides the same information on a seasonal basis. We first notice the effect of the resolution. Annual precipitation from ERA5-Land and PDIR is very smooth on the considered domain compared to the other gridded products available at a finer resolution. The precipitation difference between plains and mountains is not very prominent in the radar products SERVAT and COMEPHORE. Comparatively, the Pyrenees, the Massif Central, and the Alps receive more annual precipitation with AROME, CERRA-Land, and SPAZM. SPAZM leads to marked patterns in the alpine valleys and produces a higher correlation to altitude. AROME gives a large amount of precipitation in mountains, especially in the Massif Central. The seven gridded precipitation products show different precipitation amounts, especially in mountainous regions, due to more or less fine horizontal resolutions, different types of assimilated data, and differences in the treatment of the altitude effect.

3 Methods

Let Pobs and Pgrid denote, respectively, the observed (by rain gauges and NPT) and gridded precipitation (annual or seasonal) expressed in millimeters. Let Z indicate the altitude in meters. The indices i and j refer to stations, while the index p indicates the pixels.

3.1 Description of the relationship between precipitation and altitude

Annual and seasonal precipitation from the gridded products is extracted at station locations to evaluate the reproduction of the altitudinal variability in the precipitation. The precipitation from the NPT is only available at the annual scale. The stations are grouped by regions defined in Fig. 1a and by 300 m elevation bands b_{alt} ranging from 0–300 to 3000–3300 m. For each elevation band and region, the median of the observed ($P_{\text{obs},b_{\text{alt}}}$) and gridded ($P_{\text{grid},b_{\text{alt}}}$) precipitation

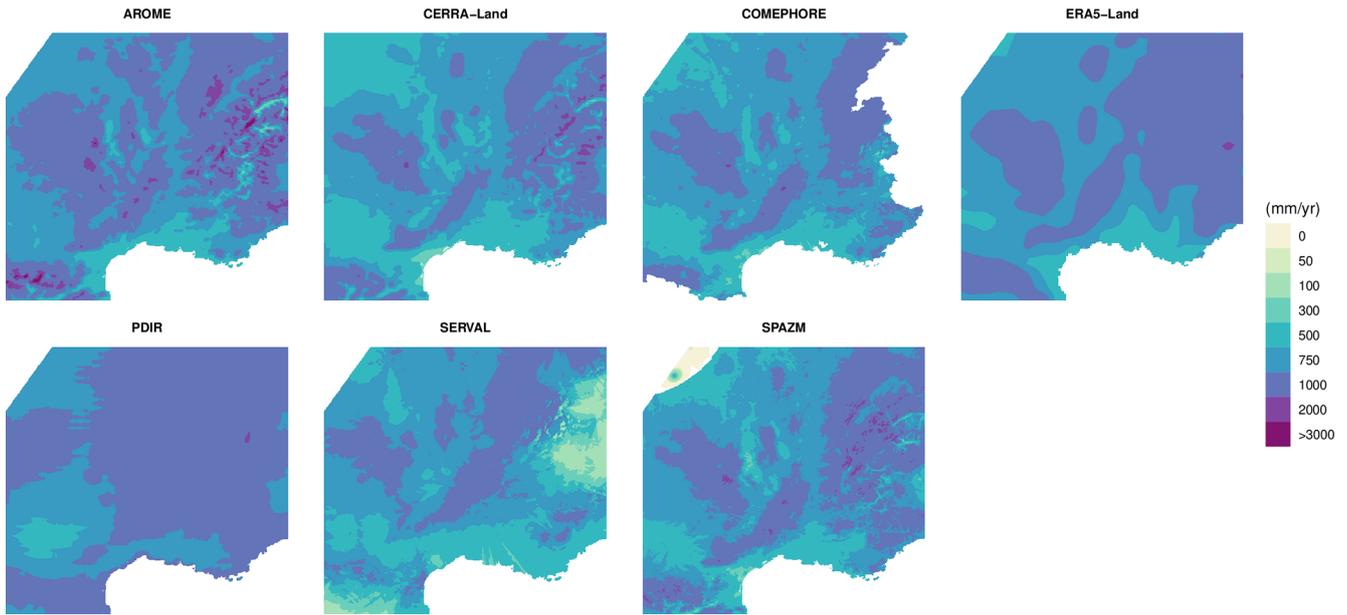


Figure 2. Mean annual precipitation for the gridded precipitation products (AROME, CERRA-Land, COMEPHORE, ERA5-Land, PDIR, SERVAL, and SPAZM) during the period 2007–2017.

(annual and seasonal) is computed using all the locations of the stations belonging to this region and elevation band:

$$P_{obs_{b_{alt}}} = \hat{P}_{obs_{i_{50\%}}}, i \in b_{alt}, \quad (1)$$

$$P_{grid_{b_{alt}}} = \hat{P}_{grid_{i_{50\%}}}, i \in b_{alt}. \quad (2)$$

The median values should represent the considered region and elevation band and be unaffected by anomalous values resulting from a specific station location. A compromise between a high number of rain gauges per band and a sufficient number of bands led to a 300 m bandwidth selection. The bands with fewer than five rain gauges are not displayed in Fig. 3a to prevent any misunderstandings or misrepresentations. To study the behavior of gridded precipitation products in high altitudes where no rain gauges are available, we compare the median of seasonal precipitation $\hat{P}_{grid_{p_{50\%}}}$, $p \in b_{alt}$ extracted from pixels within the altitudinal bands.

To test a gridded precipitation product reproducing the spatial variability in the relationship between seasonal precipitation and altitude at a finer scale, we also compute the precipitation relative difference (PRD) for several pairs of close neighboring stations with sufficient elevation difference. The two stations should have a separation distance lower than 20 km and have an elevation difference greater than 100 m. PRD should not be assimilated to precipitation lapse rates (PLRs) calculated at the catchment scale (see Sect. 3.2). For the neighboring stations i and j , we compute PRD at the station locations for observed ($PRD_{P_{obs_{ij}}}$) and

gridded ($PRD_{P_{grid_{ij}}}$) precipitation:

$$PRD_{P_{obs_{ij}}} = 100 \times 100 \times \frac{P_{obs_i} - P_{obs_j}}{Z_i - Z_j} \bigg/ \frac{P_{obs_i} + P_{obs_j}}{2}, \quad (3)$$

$$PRD_{P_{grid_{ij}}} = 100 \times 100 \times \frac{P_{grid_i} - P_{grid_j}}{Z_i - Z_j} \bigg/ \frac{P_{grid_i} + P_{grid_j}}{2}. \quad (4)$$

The gradient of seasonal precipitation relative to the altitude is divided by the mean precipitation of the two stations and multiplied one time by 100 to get the result in percent relative to the mean precipitation. This result is again multiplied by 100 to express PRD as percentages per hundred meters ($\% (100 \text{ m})^{-1}$).

3.2 Regression at nested spatial scales

The primary focus of this study is the investigation of the spatial variability in the seasonal PLRs at the sub-regional and catchment scales. It is possible to use stations at the regional scale, but the PLRs would be too dependent on their spatial sampling at the sub-regional and catchment scales. Indeed, some catchments contain zero or only one station (Fig. 1c). As an alternative, we propose using AROME to study the spatial variability in the PLR at the sub-regional and catchment scales. The choice of AROME will be motivated in Sect. 4.1. We assume linear relationships between seasonal precipitation and altitude. This strong hypothesis will be discussed later in Sect. 4.3. The PLRs are calculated with linear regression between seasonal precipitation and altitude and expressed as follows:

$$Parome_p = \alpha + \beta \times Z_p + \epsilon_p, \text{ with } \epsilon_p \sim \mathcal{N}(0, \sigma^2). \quad (5)$$

Parome_{*p*} is the seasonal precipitation (in mm) for the pixel *p*, α is the *y* intercept (in mm), β is the slope expressed (in mm m⁻¹), and Z_p is the altitude (in m) of the pixel *p*. The regression fit is performed using the maximum likelihood method implemented in the **R** function `lm`. PLRs are derived from β using the following formula:

$$\text{PLR} = 100 \times 100 \times \frac{\beta}{\bar{P}}, \quad (6)$$

where \bar{P} is the average seasonal precipitation from AROME over the area where the linear regression is performed. The slope of the regression β is multiplied by 100 twice, first to obtain the result as a percentage and once more to express PLR as a percentage per hundred meters (% (100 m)⁻¹), relative to the average seasonal precipitation from AROME. The expression of PLR (in % (100 m)⁻¹) overcomes the biased nature of gridded precipitation products and allows a spatial comparison despite significant spatial and seasonal variation in precipitation amounts across France (Fig. 2). To investigate the dependence of PLR on a given spatial scale, regressions will be conducted on 23 massifs of the Alps and catchments. On the catchments, a regression will only be performed if there is sufficient variability in elevation, i.e., if the standard deviation of altitude is higher than 50 m.

3.3 Linearity of the regressions

In our formulation (Eq. 6), PLRs are defined as the slopes of the linear regressions. The linearity of the relationship between annual precipitation and altitude at the catchment scale is a questionable hypothesis. For example, Avanzi et al. (2021) found non-linearity in two catchments located in the Italian Alps. Smoothing splines provide more flexibility than linear regressions to model annual precipitation through altitude. However, smoothing splines are not interpretable in terms of PLRs. For this reason, we use piecewise linear regressions implemented in the **R** package called `segmented` (Muggeo, 2008). We perform piecewise linear regressions (unknown break point between the first and third quartile of altitude) between annual precipitation and altitude on all catchments with sufficient variability (standard deviation of the altitude higher than 50 m) in altitude. To compare the linear (M_0 model) and the piecewise (M_1 model) linear models, taking inspiration from the Bayes factor defined in the Bayesian paradigm (Kass and Raftery, 1995), we compute the logarithm of the Bayesian information criterion (BIC) ratio (B_r) as

$$B_r = \frac{\text{BIC}_{M_0} - \text{BIC}_{M_1}}{2}. \quad (7)$$

If the logarithm of B_r equals zero, then the two models fit the data with the same quality. Negative values indicate better fits, and positive values refer to worse fits than the linear regression.

Table 2. Interpretation of the chosen categories of B_r . M_0 refers to the linear model and M_1 to the piecewise linear model.

B_r category (log scale)	Interpretation
< 0	Anecdotal evidence for M_0
0 to 4.5	Anecdotal evidence for M_1
4.5 to 9	Moderate evidence for M_1
9 to 13.5	Strong evidence for M_1
> 13.5	Decisive evidence for M_1

4 Results

4.1 Ability of gridded precipitation products to reproduce the variability in the relationship between annual/seasonal precipitation and elevation

The altitude dependencies of the gridded precipitation products are compared to those from rain gauge data in the regions presented in Fig. 1a. Figure 3a displays the median annual precipitation by 300 m altitudinal bands for the seven gridded precipitation products extracted at the location of the stations. In all regions of interest, annual precipitation observed at stations (i.e., the black curves showing the annual precipitation obtained with the rain gauges and NPTs) increases as a function of the altitude. In the mid-range mountains such as the northern Massif Central, the southern Massif Central, and the foothills of the Alps, the annual precipitation always increases from one altitudinal band to another. The maximum increase in observed annual precipitation occurs in the foothills of the Alps, with a gain of 500 mm in 1500 m altitude. In more complex topographical regions such as the Northern Alps, the Southern Alps, and the Pyrenees, the enhancement of annual precipitation with the altitude is unclear, and only visible at a reduced range of altitude. For instance, the observed annual precipitation does not seem to increase in the Pyrenees below 1800 m. In mountain regions, even the mid-range ones like the foothills of the Alps, ERA5-Land, PDIR, and SERVAl often lead to more moderate increases in annual precipitation with altitude compared to the stations. In the Cévennes/Causse region, annual precipitation from ERA5-Land slowly changes with altitude. In the northern Massif Central, all gridded precipitation products, except for PDIR, are in agreement with the observed enhancement of precipitation with altitude. In the foothills of the Alps, the underestimation of SERVAl increases through the altitudinal bands, reaching 1000 mm at 1500 m altitude. AROME overestimates annual precipitation in all regions of interest, and the bias generally increases at high altitudes. For example, at 2400–2700 m in the Southern Alps, AROME captures more than 1500 mm of the median annual precipitation, while the stations record only 1000 mm. Nevertheless, median annual precipitation from AROME and observations are strongly correlated. Without data assimilation, AROME

reproduces abrupt rises in precipitation at high altitudes. In the Northern Alps, AROME and the observed precipitation increase by the same amount of 500 mm from 1800–2100 to 2100–2400 m. In some regions of interest such as the Pyrenees and the northern Massif Central, the bias of AROME is constant through the altitudinal bands. It reflects that, in those regions, the altitude impacts AROME and the station precipitation in the same way. The reanalysis CERRA-Land, based on rain gauges, is biased in some regions (Cévennes/Causse, Northern Alps, Pyrenees, and southern Massif Central). The bias is not constant and can be positive at low altitudes and negative at high altitudes, such as in the Pyrenees region. The bias of CERRA-Land is relatively small in comparison to the bias of ERA5-Land, PDIR, SERVAL, and AROME as it reaches a maximum of 375 mm in the Northern Alps at 1800–2100 m. The reanalysis COMEPHORE and the interpolator SPAZM accurately reproduce precipitation data from stations in all regions of interest as both products assimilate rain gauge values. However, COMEPHORE underestimates precipitation above 2100 m, which is only visible in high-altitude regions, such as the Northern Alps and the Southern Alps. In the Northern Alps, the median annual precipitation from COMEPHORE is smaller than 1000 mm at 2400–2700 m, compared to more than 1300 mm measured with stations for the same altitudinal band. Figure 3a shows that COMEPHORE, CERRA-Land, and SPAZM reproduce the relationship between altitude and observed annual precipitation in the regions of interest well. AROME, without rain gauge assimilation, exhibits the same precipitation/altitude trends as the stations, although it usually overestimates annual precipitation, especially at high altitudes. It can be added that the relationship between annual precipitation and altitude derived from the location of stations is dependent on the spatial sampling, which explains the variability in the curves in Fig. 3a. Nevertheless, the poor relationships between the annual precipitation obtained with ERA5-Land, PDIR, and SERVAL and the altitude disqualify them from a further analysis of the PLRs.

Figure 3b shows the median precipitation by altitudinal bands on a seasonal basis at station locations (solid curves) and by considering all the pixels belonging to a region (dotted curves). The study of seasonal precipitation is extended above the altitude of the stations. The results are only displayed for the four highest-altitude regions (foothills of the Alps, Northern Alps, Pyrenees, and Southern Alps). Only precipitation from AROME, CERRA-Land, COMEPHORE, and SPAZM is investigated seasonally because of the highly biased nature of annual precipitation from ERA5-Land, PDIR, and SERVAL. Figure S2 in the Supplement presents the results for all seasons, products, and regions. At station locations (solid curves), the seasonal precipitation from AROME, CERRA-Land, COMEPHORE, and SPAZM is close to the observed precipitation. Only AROME overestimates seasonal precipitation, but the bias is not altitude-related. The relationship between seasonal precipitation and

altitude is different at station locations (solid curves) and with all pixels (dotted curves) in the Northern and Southern Alps. The dotted curves are more regular and above the solid ones, illustrating that the stations are not representative of the altitudinal bands of the regions. When all pixels belonging to a region (dotted curves) are taken into account, differences grow among seasonal precipitation from AROME, CERRA-Land, COMEPHORE, and SPAZM. COMEPHORE experiences both lower winter and summer precipitation than AROME, CERRA-Land, and SPAZM. COMEPHORE precipitation remains unchanged or even decreases with altitude. As a result, the differences in precipitation amounts between COMEPHORE and the other products become larger in high altitudes. For example, in the Northern Alps, the winter precipitation from AROME, CERRA-Land, and SPAZM increases by 250 mm between 3000–3300 m and 3300–3600 m, whereas COMEPHORE declines between these two altitudinal bands, leading to a 350 mm gap. CERRA-Land and SPAZM show common seasonal precipitation patterns with altitude in the Northern Alps and the Pyrenees. In the Northern Alps, CERRA-Land and SPAZM agree on the slow changes in seasonal precipitation below 3000 m and more rapid increases above this threshold. In the Southern Alps, the increase in the seasonal precipitation with altitude is larger with SPAZM than CERRA-Land. AROME produces a larger amount of seasonal precipitation in comparison to COMEPHORE, CERRA-Land, and SPAZM, especially in winter. In summer, AROME shows a suspicious precipitation amount of 1500 mm at high altitudes in the Northern Alps. In winter, the difference between AROME and the other gridded products is larger at high altitudes. For instance, in the foothills of the Alps, AROME matches SPAZM, producing 250 mm at 0–300 m, but is 300 mm higher at 2400–2700 m.

To investigate the ability of AROME, CERRA-Land, COMEPHORE, and SPAZM to reproduce the observed precipitation change with altitude at a fine spatial scale, Fig. 4 shows scatter plots of observed versus gridded precipitation relative difference (PRD). PRDs are defined for two neighboring (closest) stations as the difference in precipitation relative to the difference in altitude (see Sect. 3.2). PRDs are computed on all pairs of the closest stations within the region of interest. Figure S3 presents the PRDs scatter plots for all seasons. The same results observed in Fig. 3b appear. PRDs from COMEPHORE and SPAZM are unbiased as the points are close to the 1 : 1 lines. The variance in the PRD errors is higher with AROME and CERRA-Land. PRDs from CERRA-Land seem unbiased on average, and those from AROME are a little too high in winter. Figure 4 also reveals more difficulties for AROME, CERRA-Land, COMEPHORE, and SPAZM to reproduce observed summer than winter PRDs. PRDs with AROME are unbiased on average. It is thus possible to use AROME to model the altitude effect on seasonal precipitation at a fine spatial scale.

Figure 5 illustrates the relationship between seasonal precipitation and elevation through two transects in the

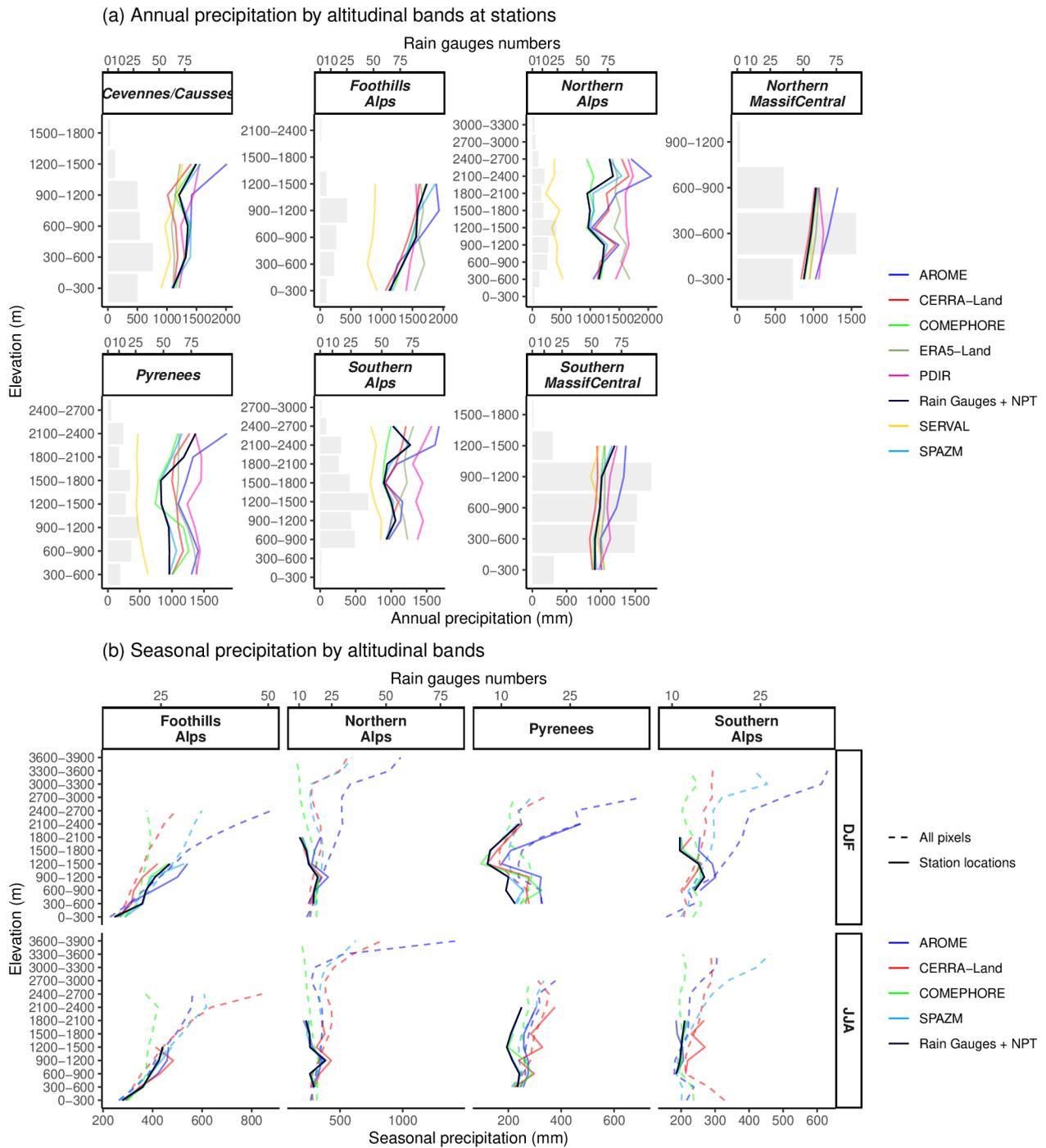


Figure 3. Comparison between the altitude dependence of precipitation from AROME, CERRA-Land, COMEPHORE, ERA5-Land, PDIR, SERVAL, SPAZM, and stations (rain gauges and NPTs) in seven regions. The lines show the median precipitation amount in each altitude zone, and the bars denote the number of rain gauges in each area. Comparisons are done at (a) the station locations at the annual scale and (b) both station locations and on all pixels from the gridded products at the seasonal scale. At the seasonal scale, only AROME, CERRA-Land, COMEPHORE, and SPAZM are analyzed for the sake of visualization. The season DJF refers to the winter (December, January, February), and the season JJA corresponds to the summer (June, July, August).

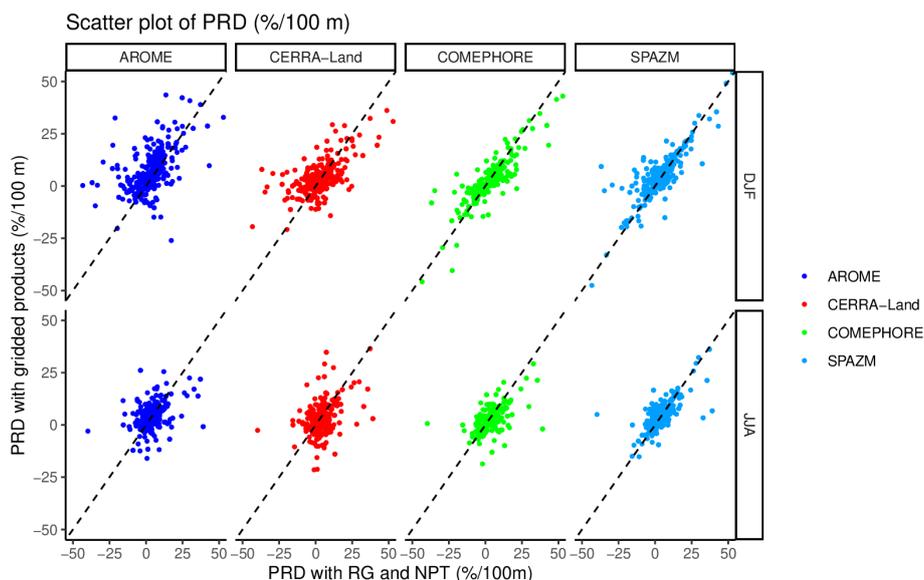


Figure 4. Comparison of precipitation relative difference ($\% (100 \text{ m})^{-1}$) computed with stations and gridded rainfall products. PRDs are derived for all pairs of neighboring stations. Scatter plot between observed and gridded PRD at station locations. The dotted lines indicate the 1 : 1 line.

Cévennes and the Northern Alps. A transect indicates a virtual line crossing the topography. Figure S4 shows the transects for all seasons. Seasonal precipitation from SPAZM is, by construction, fully correlated to the altitude, reacting to slight variations in the terrain. COMEPHORE and CERRA-Land are both less correlated to the altitude than SPAZM. CERRA-Land gives the same winter precipitation as SPAZM but different summer precipitation patterns. COMEPHORE produces similar amounts of precipitation to SPAZM in the Cévennes transect but lower ones in the Northern Alps transect. AROME has a strong relationship with the altitude. It is the gridded product with usually the highest seasonal amounts in high altitudes. For example, winter precipitation reaches 600 mm in the Cévennes with AROME, almost 500 mm with SPAZM, and 400 mm with COMEPHORE and CERRA-Land.

COMEPHORE and CERRA-Land sometimes exhibit a negative correlation to the altitude. The behavior of COMEPHORE is surprising in the Northern Alps between 8 and 15 km after the beginning of the transect. The COMEPHORE summer precipitation amount is the lowest for the highest pixel on the transect. CERRA-Land is different from the other gridded products in summer for Cévennes, with a negative association with the altitude for the first 80 km of the transect. SPAZM exhibits a higher correlation than AROME to the local topography. As a result, SPAZM precipitation maximums coincide with the summits, while those of AROME are reached before the summits in the Northern Alps, for example. Another major difference between AROME and SPAZM is the seasonality of precipitation. In the first 20 km of the Northern Alps transect,

while SPAZM shows similar seasonal precipitation patterns, AROME seasonality is more marked, with two peaks in precipitation in winter at 5 and 15 km and only one in summer. Moreover, in the Northern Alps transect, SPAZM produces higher summer precipitation than winter precipitation (400 mm against 300 mm), unlike AROME.

Overall, AROME and SPAZM demonstrate the ability to replicate the altitude-dependent variations in the annual and seasonal precipitation in the regions of interest. In comparison to commonly used gridded precipitation products such as satellite and radar data, AROME and SPAZM exhibit stronger correlations with observed precipitation. The evolution of AROME seasonal precipitation with altitude is consistent with that of stations, despite suspicious summer precipitation in very high altitudes in the Northern Alps and a too rapid rise in winter precipitation. In regions with sparse networks of stations, products based on stations such as SPAZM may be unable to capture the impact of local topography on precipitation. Since our objective is to investigate the spatial variability in the PLR at the catchment scale, SPAZM will be inadequate for this purpose. AROME should be preferred to SPAZM in ungauged areas. PLRs are therefore computed with AROME simulations from 1982 to 2018. The long temporal period of more than 30 years allows the investigation of PLR at the climatological scale.

4.2 Spatial and seasonal variability in the precipitation lapse rates

Initially, seasonal PLRs are computed at the sub-regional scale on the 23 massifs in the French Alps. Second, seasonal PLRs are computed on catchments that form a subdi-

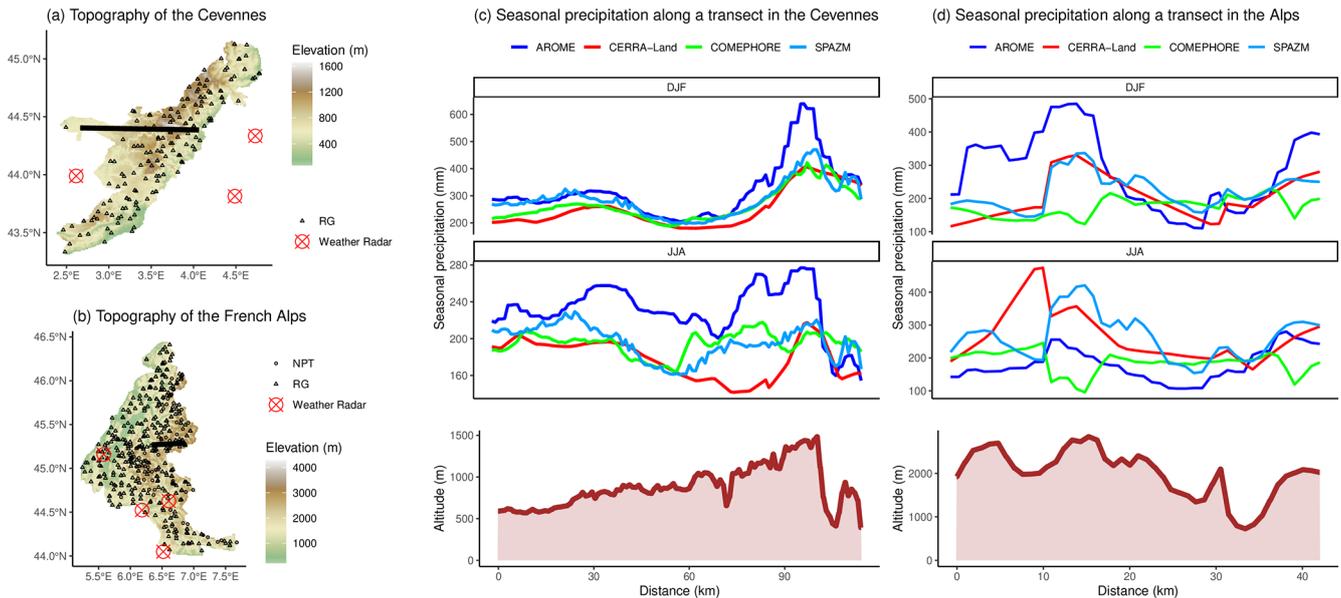


Figure 5. Transects in the Cévennes (a) and the Northern Alps (b). Winter DJF and summer JJA precipitation extracted across the transects are colored according to gridded products (c, d). Elevation is represented in brown as an altitude profile.

vision of the massifs. Figure 6a displays the R^2 at the sub-regional (massif) scale of the regressions between AROME seasonal precipitation and the altitude. Values of PLRs and R^2 at the sub-regional scale are provided in Table 4. In winter, values of R^2 are close to 0.5 in the foothills of the Alps (Chartreuse, Bauges, Aravis, and Chablais), in the Northern Alps (Beaufortain, Belledonne, and Mont Blanc), and in the Southern Alps (Queyras, Thabor, Oisans, Devoluy, Champ-saur, Pelvoux, Ubaye, and Haut Var–Haut Verdon). Regression quality in the Central Alps (Grandes Rousses, Maurienne, Vanoise, and Haute Tarentaise) is not high enough to conclude on PLR values. Winter precipitation from AROME is, for instance, not correlated to the altitude in the Haute Tarentaise massif (R^2 equals 0.12). In summer, the same spatial pattern is observed with lower values of R^2 , indicating that the altitude is better correlated with winter precipitation than summer precipitation in the French Alps.

The fine resolution of AROME provides the possibility for the investigation of the spatial variability in the relationship between seasonal precipitation and altitude at the catchment scale. Figure 6c shows the R^2 values at the catchment scale. R^2 values obtained from regressions are higher at the catchment scale than at the sub-regional scale for both winter and summer seasons. Similar to the sub-regional scale, R^2 values are generally higher in winter than in summer. In winter, the regression quality does not show a clear spatial pattern, except for lower R^2 in the eastern Alps (Haute Tarentaise, Haute Maurienne, Thabor, and Queyras). The R^2 values show a large spatial variability. In the foothills of the Alps (Vercors, Chartreuse, Bauges, and Aravis), strong R^2 (0.75–1) can be found at the borders of the massifs next to

small R^2 (0–0.5) inside the massifs. In summer, the regressions are of poor quality in the Northern and Central Alps. The massifs in the Southern Alps and some in the foothills of the Alps (Chartreuse, Bauges, and Chablais) do not show a clear deterioration of R^2 compared to the winter. In some massifs like the Maurienne in winter, the R^2 values are more than 0.5 at the catchment scale and close to 0 at the sub-regional scale. In the same way, the massifs, where the regression fits are correct at the sub-regional scale (Thabor and Haute-Maurienne), do not necessarily contain the best fit at the catchment scale. It indicates that the previous regressions (Fig. 6a) are likely conducted over excessively broad areas.

Figure 6b and d, respectively, display the PLRs of AROME seasonal precipitation at the sub-regional and catchment scales. We remark higher PLRs at the catchment scale (a mean of $5.43 \text{ } (100 \text{ m})^{-1}$ in winter and $3.31 \text{ } (100 \text{ m})^{-1}$ in summer) than at the sub-regional scale (a mean of $4.73 \text{ } (100 \text{ m})^{-1}$ in winter and $2.79 \text{ } (100 \text{ m})^{-1}$ in summer). We consider that a R^2 value smaller than 0.5 is too small to interpret the slope of the regressions. The spatial variability in the seasonal PLRs can therefore not be investigated at the sub-regional scale. We will only describe PLR values at the catchment scale because of better regressions. Figure 6d reveals higher PLRs in winter, with very high values ($8 \text{ } (100 \text{ m})^{-1}$ – $15 \text{ } (100 \text{ m})^{-1}$) and high values ($5 \text{ } (100 \text{ m})^{-1}$ – $8 \text{ } (100 \text{ m})^{-1}$) mainly located at the border of the Alps (Devoluy, Thabor, and Chablais). Moderate ($3 \text{ } (100 \text{ m})^{-1}$ – $5 \text{ } (100 \text{ m})^{-1}$) PLRs are present in almost all massifs and do not show a clear spatial pattern, such as a north–south separation. Small ($0 \text{ } (100 \text{ m})^{-1}$ – $3 \text{ } (100 \text{ m})^{-1}$) PLRs are mainly found inside the massifs of

the foothills of the Alps (Vercors, Chartreuse, and Aravis) and in the far Southern Alps (Mercantour and Haut Var–Haut Verdon). In summer, the majority of PLRs are small ($0 \text{ } (100 \text{ m})^{-1}$ – $3 \text{ } (100 \text{ m})^{-1}$), but some catchments at the border of the Alps (Bauges, Devoluy, and Haut Var–Haut Verdon) have moderate ($3 \text{ } (100 \text{ m})^{-1}$ – $5 \text{ } (100 \text{ m})^{-1}$) or high PLRs ($5 \text{ } (100 \text{ m})^{-1}$ – $8 \text{ } (100 \text{ m})^{-1}$). Figure 6 shows that the quality and slope of the regression can vary widely across the French Alps, with higher values at the borders. A sub-region like the Vercors massif covers around 1350 km^2 and presents a considerable geographic heterogeneity in PLR values. Figures S5, S6, S7, and S8 illustrate the results of the regressions for all seasons.

Figure 7 shows the winter and summer PLRs at the catchment scale in the study domain. Results for the other seasons are available in Figs. S9 and S10. The catchments located in the Morvan, the Massif Central, the Jura Mountains, the eastern Pyrenees, and the catchments close to the Mediterranean with enough altitude variation (standard deviation of the altitude higher than 50 m) have the strongest PLRs. These values are higher than those of the Alps, as they can reach $32 \text{ } (100 \text{ m})^{-1}$. Values between $8 \text{ } (100 \text{ m})^{-1}$ and $15 \text{ } (100 \text{ m})^{-1}$ are also considerably more frequent than in the Alps. In the study domain, 40 % of the catchments with $R^2 > 0.5$ have winter PLR greater than $8 \text{ } (100 \text{ m})^{-1}$. This percentage drops to 15 % if we only consider the Alps. The slopes of the regression have higher variability in the Alps and the Pyrenees. PLRs are seasonally varying, with generally the highest values in winter. In winter, the Massif Central hosts the strongest PLRs. In summer, the Mediterranean area, ranging from the eastern Pyrenees to the Mediterranean Alps, has the highest PLRs.

To summarize, PLRs are generally higher in winter than in summer. PLRs are sensitive to the spatial scale, with higher values at the catchment scale than the sub-regional (massif) scale. Spatial patterns of significant PLRs vary according to the season in France. In winter, the higher PLRs reach $32 \text{ } (100 \text{ m})^{-1}$ and are mainly located in the Massif Central. In summer, PLRs are lower and reach larger values on the Mediterranean coast. PLRs vary notably in space, even within a single mountain range such as the French Alps.

4.3 Linearity of the relationship between annual precipitation and altitude

To assess the linearity of the regressions performed in Fig. 7, piecewise linear regressions were compared to linear regressions through the BIC ratio (see Sect. 3.3). Figure 8a illustrates the BIC ratio categories (Table 2) with some examples of scatter plots and corresponding regressions. Colors refer to categories of the logarithm of B_r . Based on the scatter plots, we consider the relationship between annual precipitation and altitude not linear when there is strong or decisive evidence for M_1 (red color). Figure 8b shows the spatial distribution of the logarithm of the B_r values. In total,

10 % of the catchments show decisive evidence for M_1 . It appears mainly in the Massif Central, the Jura Mountains, and the Southern Alps. In the other regions (Northern Alps and Pyrenees), it is sufficient to use a linear regression to derive PLR. Figure 8c shows the distribution in surface areas of the catchments among the five categories of B_r . Strong B_r are mainly found for large catchments, revealing the spatial-scale dependence of PLR discussed in Sect. 5.3. Too much spatial heterogeneity is still present in some catchments (last scatter plot in Fig. 8a), suggesting an even finer spatial division and taking into account the aspect and the slope of the terrain faces. Large catchments can have slopes with different orientations, for instance. Figure S12 provides the same diagnostic of the linearity for AROME seasonal precipitation.

In 90 % of the catchments, the regression between seasonal precipitation and altitude can be considered linear. The remaining 10 % often corresponds to larger catchments which could motivate the estimation of the PLRs on smaller areas in these cases.

5 Discussion

5.1 Ability of gridded precipitation products to capture the relationship between annual/seasonal precipitation and altitude

This article reviews the potential of seven gridded precipitation products for precipitation gradient estimation in a complex topographic region. ERA5-Land, PDIR, and SERVAl underestimate annual precipitation gradients even for mid-range mountains. The limitations of these three gridded precipitation products are even more apparent in high-altitude regions such as the Northern Alps and the Southern Alps. The assimilation of satellite data and the non-use of rain gauge values in ERA5-Land and PDIR are an obstacle to their utilization in a complex topographic region such as France. These findings converge with those of Jiang et al. (2022), where they found that the satellite product Information Measures for the EU Cohesion policy (IMREG) cannot reproduce the precipitation gradient in Tibet. SERVAl does not use altitude as a co-variable to determine precipitation from individual radars. The uniform ratio applied for this product in the hourly correction only represents the station elevation range and not the entire radar coverage. The translation of reflectivity to precipitation amount using the vertical profile of reflectivity is another potential source of error that results in inaccurate negative precipitation gradients in Fig. 3. In their study of the French Alps at the sub-regional level, Faure et al. (2019) found a significant underestimation of precipitation gradient using PANTHERE, SERVAl's predecessor.

CERRA-Land, COMEPHORE, and SPAZM assimilate ground precipitation measurements from weather stations and are thus correlated with the annual precipitation observed

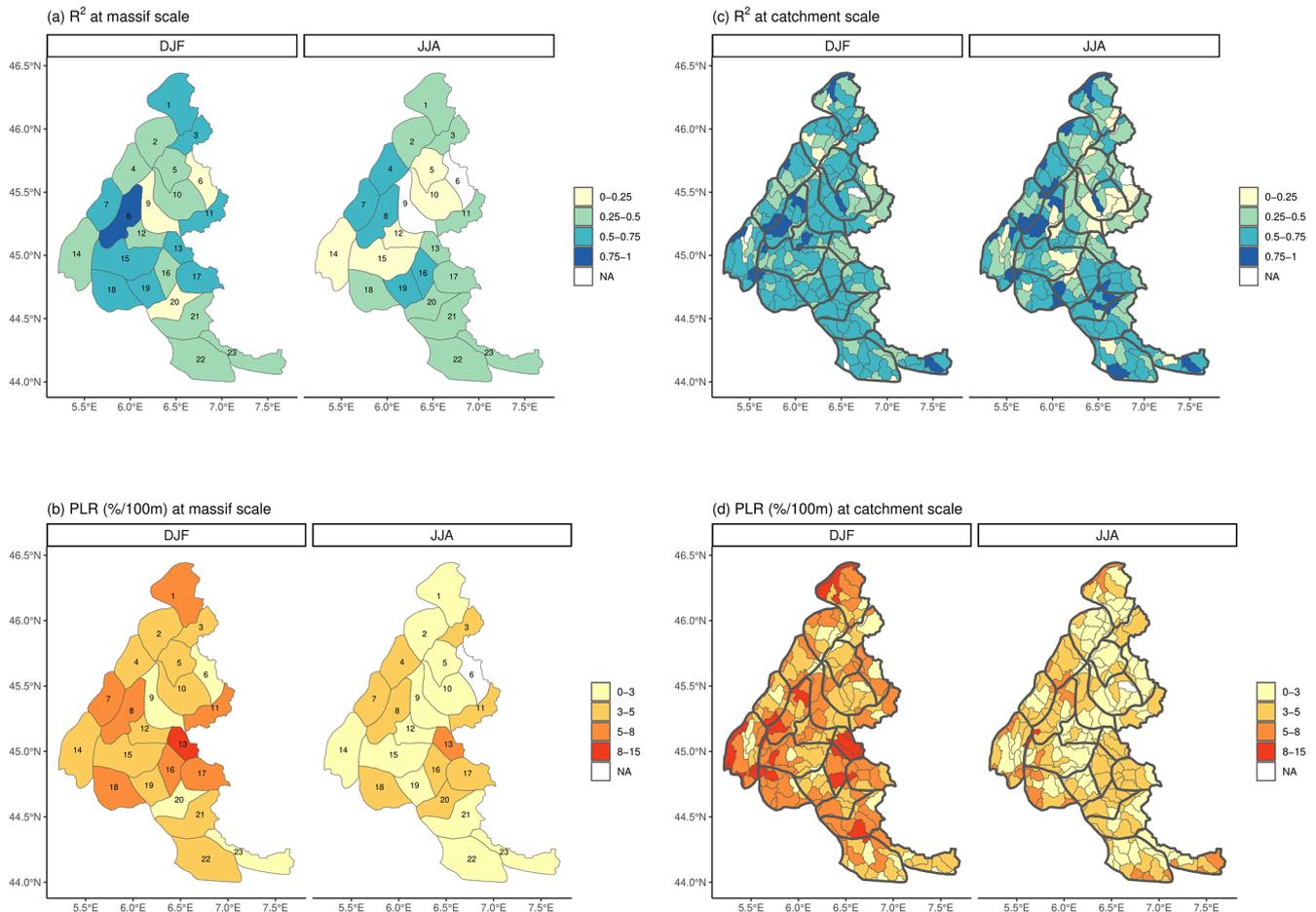


Figure 6. Spatial distribution of R^2 (a, c) and precipitation lapse rates (precipitation change per 100 m altitude difference divided by area-averaged precipitation) (b, d) extracted from the precipitation–altitude regressions for all grids within each area. Categorical results are displayed in the French Alps at regional and catchment scales for both winter DJF and summer JJA precipitation. The area indicated as “NA” and filled in white represents regions with insufficient altitudinal variability to compute precipitation gradient. The indexes of the massifs are also printed and refer to Table 3.

at the stations. COMEPHORE leads to an adequate relationship between annual/seasonal precipitation and altitude in mid-mountain ranges highly instrumented as the southern Massif Central. However, in more complex topographical regions with sparse stations such as the Northern Alps, COMEPHORE relies heavily on radar data from high altitudes and is unable to accurately reproduce the increase in the precipitation with altitude. COMEPHORE processes fewer high-altitude stations than SPAZM, making it unable to represent precipitation in high-altitude areas correctly. The sharp decrease in the summer precipitation with altitude from COMEPHORE in the Northern Alps illustrated in Fig. 5d is likely due to the shielding of the radar beams by mountain ranges (Germann et al., 2006). CERRA-Land can replicate the observed seasonal precipitation enhancement with the altitude. However, at the annual scale, CERRA-Land is slightly biased in some regions, and the bias may change the sign according to altitudinal bands such as in the Pyrenees

(Fig. 3a). SPAZM can reproduce the relationship between annual precipitation at station locations and altitude at the regional scale. This result is not surprising, as SPAZM interpolates precipitation observations from the stations, incorporating the effect of the altitude through local linear regressions. However, SPAZM is affected by the density of stations. In regions where the stations are more sparse, as in the Pyrenees, the annual precipitation from SPAZM does not match the annual precipitation from the stations. This suggests a possible limitation of SPAZM in ungauged mountainous regions. AROME, without rain gauge assimilation, can reproduce the relationship between seasonal precipitation and altitude despite a positive bias in winter in mountainous regions. We find that AROME produces higher annual/seasonal precipitation than the other gridded products. The difference in seasonal precipitation is limited at low altitudes and rises sharply at high altitudes. Higher precipitation accumulations with RCMs have been documented. In western Montana, in

Table 3. Values of precipitation lapse rates (precipitation change per 100 m altitude difference divided by area-averaged precipitation) and R^2 at the sub-regional scale for the 23 French massifs for both winter DJF and summer JJA. The Massif number is indicated in Fig. 6.

Massif number	Massif name	DJF PLR (%/100 m)	JJA PLR (%/100 m)	DJF R^2	JJA R^2
1	Chablais	5.9	2.7	0.57	0.45
2	Aravis	4.3	2.3	0.45	0.43
3	Mont Blanc	4.1	4.3	0.66	0.45
4	Bauges	4	3.6	0.5	0.52
5	Beaufortain	4.1	1.9	0.45	0.21
6	Haute Tarentaise	2.2	0	0.11	0
7	Chartreuse	5.4	3.9	0.57	0.58
8	Belledonne	6.6	4.1	0.8	0.66
9	Maurienne	2.1	0.3	0.11	0
10	Vanoise	3.5	0.6	0.31	0.01
11	Haute Maurienne	5.2	3.5	0.51	0.39
12	Grandes Rousses	3.6	1.3	0.28	0.09
13	Thabor	9.1	5.2	0.52	0.34
14	Vercors	4.5	1.7	0.31	0.14
15	Oisans	4	1.4	0.55	0.21
16	Pelvoux	6.3	4.3	0.49	0.52
17	Queyras	5.5	4.2	0.51	0.49
18	Devoluy	6.8	3.5	0.56	0.3
19	Champsaur	4.7	3	0.67	0.58
20	Parpaillon	2.8	3.2	0.17	0.5
21	Ubaye	4.3	2.2	0.48	0.46
22	Haut Var–Haut Verdon	3.2	2.3	0.49	0.36
23	Mercantour	2.6	2.9	0.28	0.33

high-altitude areas where observations are sparse, Silverman et al. (2013) noted a higher annual precipitation with the Weather Research and Forecasting (WRF) model than with the Parameter-elevation Regressions on Independent Slopes Model (PRISM) (Daly et al., 1994). In India, Li et al. (2017) found that annual precipitation obtained with WRF is double or triple the amount of the satellite precipitation product Tropical Rainfall Measuring Mission (TRMM). The correlation between AROME and COMEPHORE on the Cévennes transect in summer (Fig. 5c) is comforting, as the radars operate well in summer in the Cévennes region (Delrieu et al., 2013). Moreover, the seasonality of the precipitation is taken into account with AROME (see Fig. 5c). AROME simulates precipitation without using rain gauge data and poorly matches the high-elevation gauges, which could be the combined effect of precipitation undercatch (Groisman and Legates, 1994; Sevruk, 1997; Pollock et al., 2018) and some deficiencies in AROME (Monteiro et al., 2022). In Fig. 3b, AROME shows a strong agreement with the rain gauges in summer (even the high-altitude ones). The differences are larger in winter, where precipitation mostly fall as snow. Precipitation undercatch is limited in summer and more important in winter with solid precipitation. This suggests that the precipitation undercatch could partly explain the differences between AROME and rain gauges.

To summarize, this study shows that the simulations of the CP-RCM AROME, despite winter precipitation overestimation, offer the opportunity to derive PLRs at a fine spatial scale. The other grid precipitation products obtained from radar, satellite, or rain gauge data show important limitations in a complex topography region. Radar is among the other products subjected to beam blocking. The satellite resolution is too coarse to represent the altitude effect on precipitation. Rain gauge networks are too sparse, and even high-density ones do not account for the spatial and altitudinal variability in the seasonal precipitation.

5.2 Values and spatial variability in the precipitation lapse rates

Table 4 synthesizes the main research on PLRs. Like Ménégoz et al. (2020) in Switzerland, we find higher PLRs in winter, likely due to the large-scale circulation of air masses. In winter, the westerly wind coming from the Atlantic loaded with moisture is dominant and hits the primary mountainous regions on its way. The affected catchments correspond to the first orographical barrier and are located in the Massif Central, extending from the Languedoc to the Morvan regions. For those catchments, 50 % of the PLRs are in the range of $5 \text{ \% } (100 \text{ m})^{-1}$ – $13 \text{ \% } (100 \text{ m})^{-1}$. For the same reason, the PLRs are in the range of $5 \text{ \% } (100 \text{ m})^{-1}$ –

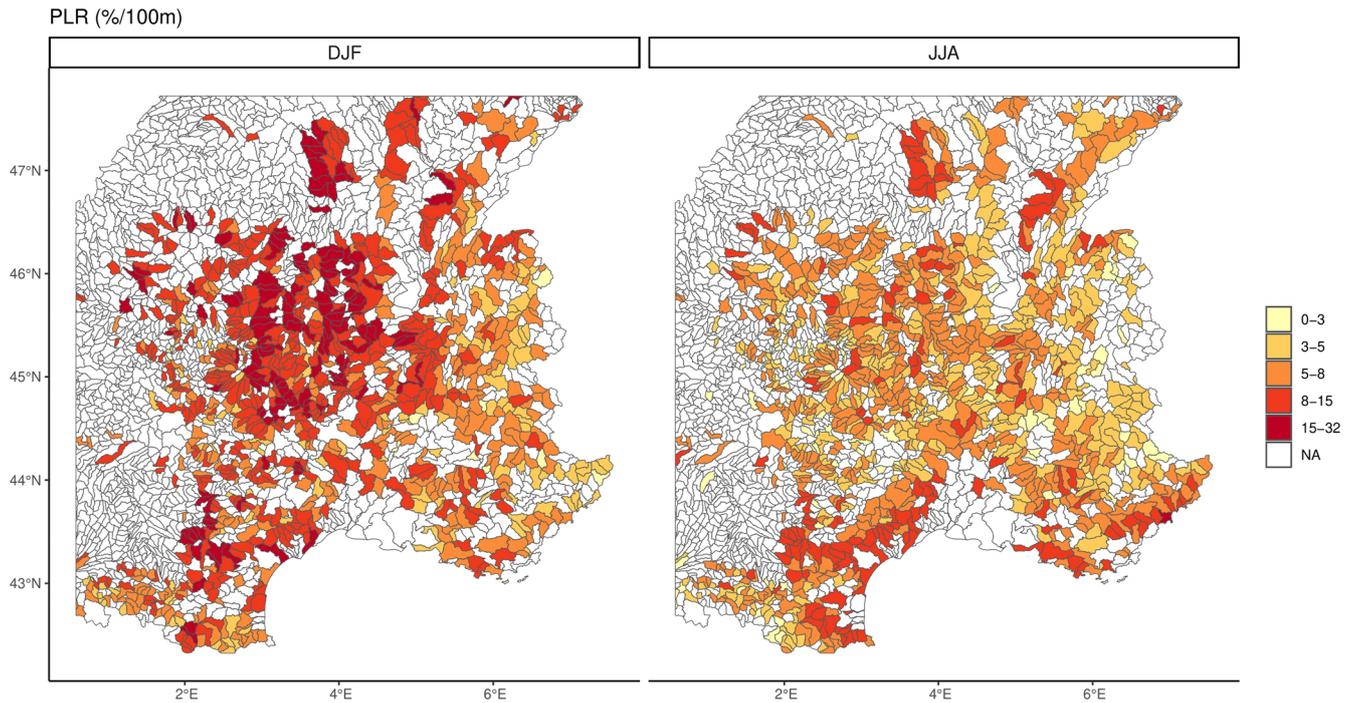


Figure 7. Spatial distribution of winter DJF and summer JJA precipitation lapse rates (precipitation change per 100 m altitude difference divided by area-averaged precipitation) extracted from the precipitation–altitude regressions for all grids within each area. Categorical results are displayed in the study domain at the catchment scale. The area indicated as “NA” and filled in white represents catchments with insufficient altitudinal variability (standard deviation of the altitude lower than 50 m) or R^2 smaller than 0.5.

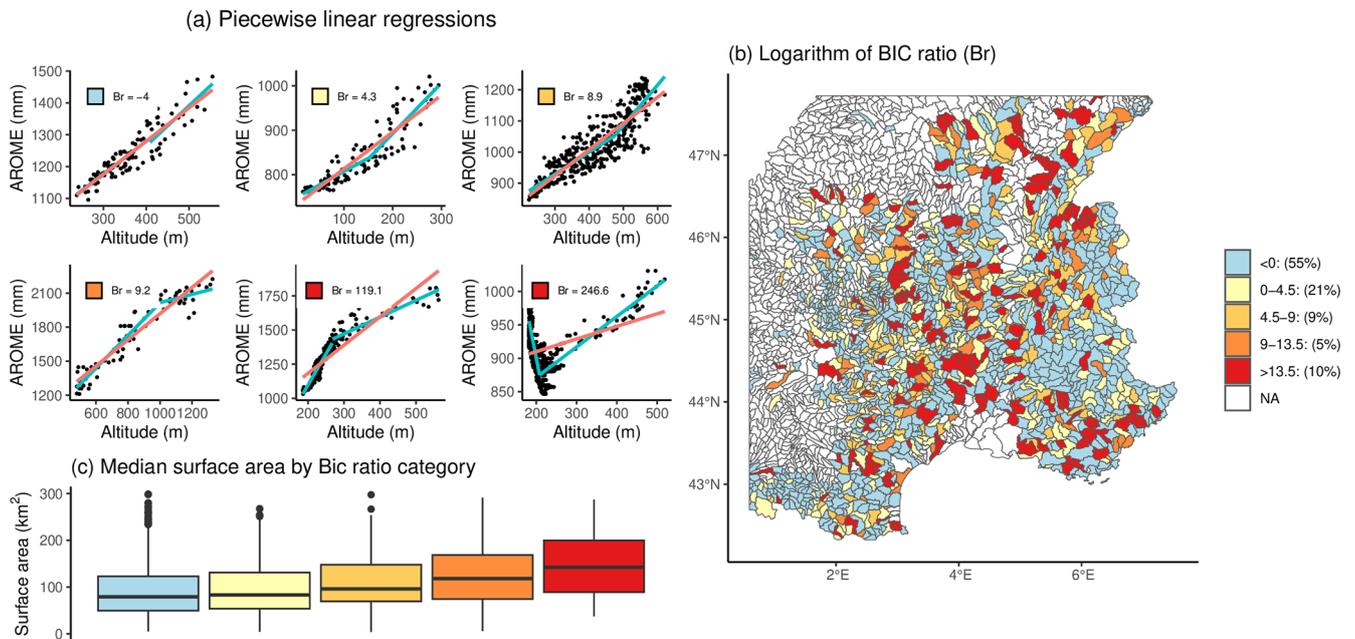


Figure 8. (a) Scatter plots of AROME annual precipitation (mm) as a function of the altitude for some catchments taken as illustrative examples of the different categories of BIC ratio. Linear (red lines) and piecewise linear (blue curves) regressions are superimposed. (b) Spatial representation of the logarithm of the BIC ratios between linear and piecewise linear regression (annual precipitation \sim altitude) models. The numbers expressed in percent correspond to the percentage of catchments within the classes of the BIC ratio. (c) Distribution of the surface area of the catchments according to the categories of BIC ratio.

15 % (100 m)⁻¹ for the first western catchments of the Alps. Interestingly, a distinct pattern emerges in the Thabor, where some PLRs are in the range of 8 % (100 m)⁻¹–15 % (100 m)⁻¹ due to frequent easterly weather fronts brought by Mediterranean circulation from Italy. In summer, the Mediterranean area is subjected to the dominant southern flows and hosts the highest PLRs. The southern air masses absorb the humidity from the Mediterranean Sea. For this reason, we found higher PLRs in the eastern Pyrenees (8 % (100 m)⁻¹–15 % (100 m)⁻¹) compared to the Central Pyrenees (3 % (100 m)⁻¹–8 % (100 m)⁻¹). Convective processes are more frequent in summer than in winter. Daily precipitation amounts from convective processes are often negatively correlated to the altitude (Schäppi, 2013). That is likely to explain the lower PLRs in summer than in winter. To summarize, we notice high PLRs in areas subjected to the prevailing winds in France, which are west, south, and east winds, depending on the season. Like Kotlarski et al. (2012), we find lower PLRs in the Alps than in the rest of France.

In the French Alps, a complex topographical region, we do not find a significant difference in PLR values between the Northern and the Southern Alps for the winter and summer seasons, contrary to Sevruk (1997), Durand et al. (2009), and Ménégoz et al. (2020). That is probably due to the finer spatial scale used in the study. At a larger spatial scale, large-scale circulation, rather than the local topography, drives seasonal precipitation amounts (Jiang et al., 2022). We find a large spatial variability in the PLRs, which can vary inside the same massif from one catchment to another. For instance, winter PLRs range from 3 % (100 m)⁻¹–5 % (100 m)⁻¹ to 8 % (100 m)⁻¹–15 % (100 m)⁻¹ in Belledonne. Using rain gauges, Ogrin and Kozamernik (2018) highlighted the extreme spatial variability in the PLRs in the range of 5 % (100 m)⁻¹–16 % (100 m)⁻¹ for the summer season in three nearby alpine valleys of Slovenia. We find the same magnitude in PLR values and variability. PLRs are highly spatially varying in high-altitude and complex topographical regions such as the French Alps and Pyrenees. In the Massif Central, PLRs exhibit less spatial variability.

In summary, PLRs are probably higher in winter because of the omnipresence of stratiform events that cause structured orographic precipitation in catchments exposed to oceanic weather patterns. In summer, the air masses are coming from the Mediterranean Sea, causing high PLRs in the coastal and hilly Mediterranean areas. In this season, precipitation is mainly caused by the combination of multiple convective events, resulting in unclear PLR spatial patterns.

5.3 Dependence of the precipitation lapse rates to the spatial scale

The low R^2 found in Fig. 6a is a reminder of the high intra-massif variability in the relationship between seasonal precipitation and altitude in large sub-regions. The sub-region dissection into more homogeneous areas, such as small

catchments with a surface area close to 100 km², allows a better description of the PLRs. Regressions at the catchment scale have larger PLR and R^2 values than at the sub-regional scale. In winter in Maurienne, all catchments have R^2 greater than 0.5, with PLRs in the range 3 % (100 m)⁻¹–8 % (100 m)⁻¹. The R^2 is close to 0, and the slope is close to 2 % (100 m)⁻¹, with a single regression covering all the Maurienne massif. Figure 9 shows the scatter plot between AROME annual precipitation and altitude for the pixels within three groups of catchments in the Vercors, a massif covering 1350 km². The colors of the pixels relate to distinct geographical locations. Figure S11 provides the scatter plots for seasonal precipitation. Figure 9 suggests that Vercors encompasses several sub-regions with different relationships between annual precipitation and altitude. This relationship is linear for the southern and western Vercors catchments. However, the western and southern Vercors are not subject to the same meteorological influences. The PLRs are higher in the western Vercors because of the dominant rainy westerly flow. The relationship is not linear for the catchments located in the Vercors plateaus. The attenuation of orographic precipitation after crossing the first hills of the west of the massif is a possible explanation. Figure 9 reveals the coexistence of PLR spatial patterns within a single massif.

In the French Alps, a complex topographical region, it is not appropriate to calculate PLRs on sub-regions that are too vast with surface areas of more than 1000 km². In vast sub-regions, there is a risk of mixing areas with too much spatial heterogeneity and losing information about the PLR spatial variability. An investigation of the PLR with stations or station-based precipitation products such as SPAZM should be conducted carefully in sparse station networks. For example, in the Aravis massif covering 1070 km², linear regressions between precipitation and altitude in SPAZM are performed using all stations within the massif to ensure robustness. However, Fig. 6d indicates that the Aravis has high intra-massif variability in the winter PLR with lower values in its western part. SPAZM cannot capture this variability.

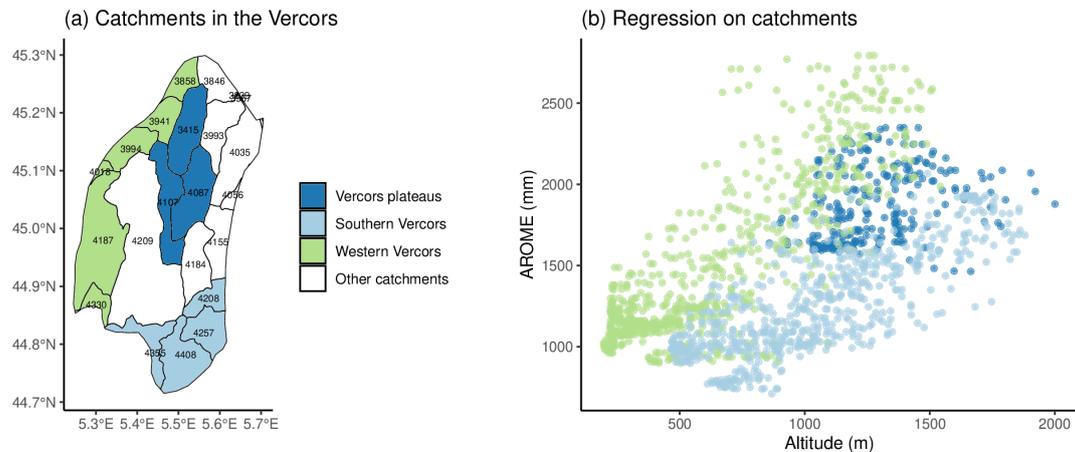
PLRs are thus scale-dependent. The linear regressions give better R^2 and larger PLRs at the catchment scale (100 km²) than at the sub-regional scale (1000 km²). Sub-regions that are too broad often hide distinct altitude effects on precipitation.

5.4 Uncertainty in the study and outlook

A limitation of our study is the potential undercatch of precipitation by the high-altitude stations. The underestimation can be as high as 50 % in winter during meteorological events with strong winds and a large proportion of snowfall. A 25 % underestimation of mountain annual precipitation is possible (Sevruk, 1997). In low altitudes, the underestimation is less severe. As a result, the comparison of PLRs in Fig. 3a can lead to wrong conclusions based on the bias of the gridded precipitation products. Despite the measuring uncertainty, we

Table 4. Summary of the studies on precipitation lapse rates (PLRs), namely study domain, data, spatial scale, period, findings, and reference.

Study domain	Data	Spatial scale	Period	Findings	Reference
Switzerland	Rain gauges	Massifs 4000 km ²	1951–1980	Northern Alps: 5.4 % (100 m) ⁻¹ Southern Alps: 0.24 % (100 m) ⁻¹	Sevruk (1997)
French Alps	SAFRAN	Regions 1000–10 000 km ²	1958–2002	Northern Alps: 29.4 mm (100 m) ⁻¹ Central Alps: 19.5 mm (100 m) ⁻¹ Southern Alps: 17.2 mm (100 m) ⁻¹ Mediterranean Alps : 17.8 mm (100 m) ⁻¹	Durand et al. (2009)
Europe	RCM	Country 200 000–1 000 000 km ²	1961–2000	Spain: -3.7 mm (100 m) ⁻¹ United Kingdom: 131 mm (100 m) ⁻¹ France (without the Alps and the Pyrenees): 32 mm (100 m) ⁻¹ Alps: 14.6 mm (100 m) ⁻¹	Kotlarski et al. (2012)
Slovenia	Rain gauges		2012–2015 (MJJASON)	5 % (100 m) ⁻¹ –16 % (100 m) ⁻¹	Ogrin and Kozamernik (2018)
Swiss and French Alps	RCM	Regions 10 000–50 000 km ²	1971–2008	French Northern Alps: 2.6 % (100 m) ⁻¹ French Southern Alps: 1.6 % (100 m) ⁻¹ Swiss Alps: 4.3 % (100 m) ⁻¹ in winter (DJF) 2 % (100 m) ⁻¹ in summer (JJA)	Ménégoz et al. (2020)
Third Pole	WRF	Regions (1 000 000 km ²) catchments (10 000 km ²)	1980–2018	Regions: 2.90 % (100 m) ⁻¹ to 11.26 % (100 m) ⁻¹ Large catchments: -10 % (100 m) ⁻¹ to 10 % (100 m) ⁻¹ Small catchments: -4 % (100 m) ⁻¹ to 10 % (100 m) ⁻¹	Jiang et al. (2022)

**Figure 9.** (a) Catchments located in the Vercors massif and colored according to their geographic locations. Three groups of catchments are distinguished: the Vercors plateaus, the western Vercors, and the southern Vercors. ID numbers are printed for each catchment. (b) Scatter plot of AROME annual precipitation and altitude for the pixels in the Vercors (gray dots). The colors of dots correspond to the colors used in panel (a).

decided to include NPTs in our study to account for high ungauged altitude regions. The available period from NPTs differs from those of rain gauges and precipitation products. Additional data sources can enhance the spatial and altitudinal representation of PLR. High-altitude measurements of snow water equivalent (SWE) levels can constrain the regressions between winter precipitation and altitude and make PLRs more robust. However, the transition from daily SWE to seasonal precipitation is highly temperature- and snowmelt-dependent. For high-altitude measurements, it is possible to make the strong assumption that only solid precipitation has

occurred and that snowmelt has been minimal, as stated in Avanzi et al. (2021). Due to the high level of uncertainty in the measure, we opt not to include those data.

At high altitudes, winter precipitation from AROME is significantly higher than that of the other precipitation products, including SPAZM. At 2700–3000 m in the Pyrenees, AROME simulates an extra 400 mm of precipitation compared to COMEPHORE, CERRA-Land, and SPAZM. SPAZM may underestimate the high-altitude precipitation due to the precipitation undercatch. It is also likely that AROME overestimates seasonal precipitation. Indeed, Mon-

teiro et al. (2022) found excessive snow accumulation with AROME in the French Alps above 1800 m. The authors advanced assumptions about spurious snow accumulation, such as the AROME cold temperature bias and the weakness of the AROME snowpack schemes. The winter PLR values given in Fig. 7b for high-altitude catchments should be considered with caution. In addition to AROME, various CP-RCMs listed in Ban et al. (2021, Table 1) have been used to simulate precipitation at a fine resolution in France. PLRs derived as the slope of linear regressions are necessarily uncertain because of the regression uncertainty. A perspective of the study could be to include prediction intervals, i.e., a range of values containing the “true” value with a probability of 0.95, for example.

6 Conclusion

We address the question of the spatial variability in the seasonal precipitation lapse rates in France, a country with varied and complex topography. A dense rain gauge network enriched with snow rain gauge totalizers characterizes the relationship between annual/seasonal precipitation and altitude in large regions. The ability to reproduce this link obtained from ground measurements is analyzed using seven different kinds of gridded precipitation products, namely ERA5-Land, the PDIR satellite product, the raw radar product SERVAl, a 37-year simulation from the CP-RCM AROME, two reanalysis products named CERRA-Land and COMEPHORE, and the precipitation interpolator SPAZM. Precipitation products commonly used in hydrology are limited in high-altitude regions with complex topography, such as the French Alps. Radar products perform well in hilly terrain but underestimate annual precipitation in high altitudes, resulting in a negative association between annual precipitation and altitude. Satellite products are not subject to the same estimation errors as radar products. However, the coarse resolutions do not accurately represent the annual precipitation altitude dependence in regions with topography heterogeneity. The performance of station-based products is dependent on the density of the weather station network. In ungauged mountainous areas, precipitation estimation from those kinds of products requires some degree of interpolation, whereas AROME uses physical laws. AROME does not incorporate rain gauge data and is therefore not subjected to bias caused by the undersampling of particular high-altitude spatial areas. The simulations from the CP-RCM AROME, despite an overall overestimation of high-altitude precipitation, offer an opportunity to assess the altitude dependence of seasonal precipitation at the catchment scale. We employ the CP-RCM AROME to derive seasonal PLRs on 23 French Alps massifs and 2748 catchments. The spatial-scale dependence of PLRs is investigated in the French Alps by comparing seasonal PLR values at the sub-regional (1000 km²) and the catchment (100 km²) scales. PLRs are derived using linear

regression between seasonal precipitation and altitude. PLRs are spatial-scale-dependent as they tend to be higher and better represented when computed at the catchment scale. The local topography influence on seasonal precipitation is major at a small spatial scale but dominated by the large-scale atmospheric circulation influence at the larger sub-regional spatial scale. PLRs are higher in winter. The majority of winter PLRs are positive. In total, 95 % of them are in the range 0.55 % (100 m)⁻¹–13.10 % (100 m)⁻¹. The higher values are found in the westerly and Mediterranean-dominant flow-exposed regions, reaching up to 32 % (100 m)⁻¹. In high-altitude regions such as the French Alps and the Pyrenees, we notice sheltering effects with PLRs close to 0 % (100 m)⁻¹ for some catchments enclosed by mountains. It is generally the first reliefs encountered by air masses that host the stronger PLRs. This article emphasizes the importance of considering the topography on a fine spatial scale to estimate PLRs in ungauged mountainous regions. CP-RCM models offer an opportunity to compute the enhancement of seasonal precipitation with the altitude.

Code and data availability. SERVAl and COMEPHORE are available from the AERIS portal (<https://radarsmf.aeris-data.fr/>, AERIS portal, 2023). PDIR is available from the CHRS Data Portal (<https://chrsdata.eng.uci.edu/>, CHRS Data Portal, 2023). ERA5-Land and CERRA-Land are available from Copernicus (<https://cds.climate.copernicus.eu>, Copernicus, 2023). AROME is available from the Med-CORDEX portal (<https://www.medcordex.eu/search/index.php>, Med-CORDEX, 2023). The spatial data set containing the catchments is available from <https://www.sandre.eaufrance.fr/atlas/atlas/fre/catalog.search#/metadata/6571f89a-0608-49a2-9ab5-f005d842e81f> (Sandre, 2020). SPAZM and the stations are not publicly available. The  codes used to perform the analysis are available upon reasonable request to the corresponding author.

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Author contributions. GE and VD conceived the idea. VD carried out the analysis and wrote the article. GE, DP, and ACF participated in the discussion and design of this study and contributed to writing and editing the paper.

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