SLOCLIM: A high-resolution daily gridded precipitation and temperature dataset for Slovenia

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Abstract. We present a new publicly available daily gridded dataset of maximum and minimum temperature and precipitation data covering the whole territory of Slovenia from 1950 to 2018. It represents the great variability of climate at the crossroads between the Mediterranean, Alpine and continental climatic regimes with altitudes between 0 – 2864 m a.s.l. We completely reconstructed (quality control and gap filling) the data for the three variables from 174 observatories (climatological, precipitation and automatic stations) with the original records all over the country. A comprehensive quality control process based on the spatial coherence of the data was applied to the original dataset, and the missing values were estimated for each day and location independently. Using the filled data series, a grid of 1x1 km spatial resolution with 20,998 points was created by estimating daily temperatures (minimum and maximum) and precipitation, and their corresponding uncertainties at each grid point. In order to show the potential applications, four daily temperature indices and two on precipitation were calculated to describe the spatial distribution of: 1) the absolute maximum and minimum temperature; 2) the number of frost days; 3) the number of summer days; 4) the intensity of precipitation; and 5) the maximum number of consecutive dry days. The use of all the available information, the complete quality control and the high spatial resolution of the grid allowed for an accurate estimate of precipitation and temperature that represents a precise spatial and temporal distribution of daily temperatures and precipitation in Slovenia. The SLOCLIM dataset is publicly available at https://doi.org/10.5281/zenodo.4108543 and can be cited as Škrk et al. (2020).

1 Introduction

Complex territories demand high-quality climatic data to properly describe and understand those events that are usually masked by global datasets or others at coarser resolutions than those required by the high-frequency climatic phenomena, like extreme events. Slovenia is an outstanding case study of a challenging territory for climate research. The high heterogeneity of the

terrestrial surfaces and the greatly varied climatic regime in a 20,271 km² country represented by lowlands and mountains within an altitudinal range of nearly 3,000 meters, serves as example of extreme climatic diversity in a relatively small region. Continental, Alpine and sub-Mediterranean climates meet and interact in the area, including two climatic zones that are

35   particularly affected by global warming (the Mediterranean and Alpine regions) (Giorgi, 2006; Klanderud and Totland, 2005). Slovenia also faces a high frequency and magnitude of extreme events (Cegnar, 2016), that are reflected at an annual scale where, for instance, precipitation totals over 3500 mm can be locally measured in the northwest, while less than 800 mm can be found in the northeast. Moreover, temperature ranges from -35 to +40 °C (Bertalanič et al., 2006) show the very extreme character of the seasons. These extremes affect the environment and life of the population through damage to infrastructure,

40   agricultural and afforested areas. More in-depth knowledge of climatic conditions is needed to design prevention measures in this context.

The Environmental Agency of the Republic of Slovenia (ARSO) maintains a very good archive of meteorological data (ARSO, 2020) collected through meteorological stations covering the whole territory and all the different environments. However, the density of these observations varies in space and time, as in most European countries, with a lack of meteorological stations

45   in sparsely populated areas including mountains (especially at higher altitudes), agricultural and afforested areas, which represent most of the Slovenian territory. Furthermore, the data provided by the stations is often of varying length and partly incomplete due for various reasons, such as: measurement errors, mechanical failure of devices, missing and interrupted measurements, abandonment and relocation of meteorological stations, amongst others. In addition, considering the complex orography and proximity to the coast, the heterogeneity of terrestrial surfaces, where weather conditions change rapidly, a

50   dense dataset with an integrated approach is needed for the whole country.

High-resolution gridded datasets allow for precise characterization of the climatic variables and are important for climatic analysis and hydrological hazard predictions, agricultural management, forestry and many other applications related to climate variables, as they help to identify extreme climatic conditions that can be analysed and used for prevention measures (Herrera et al., 2012). In this sense, gridded datasets have an advantage over station-based datasets, as they can provide climatic data

55   for any location in the territory in the form of serially complete data series that allow long-term analysis. They allow us to examine spatial and temporal climatic variations providing an equal area grid where each grid value is under validation control (Haylock et al., 2008), and missing information from stations is reconstructed. On the other hand, they have some limitations, as they are models and not real measurements. How much they differ from real data depends on the accuracy of the method and locations/density of the meteorological stations. Furthermore, the reconstruction of long periods with missing data can

60   lead to inhomogeneities in the temporal structure of the reconstructed series. When the temporal scale is increased from monthly to daily data, the spatial and temporal variability is also increased, and the uncertainty can quickly escalate.

With regard to Slovenia, and to the best of our knowledge, until recently the only gridded climate dataset was the one created by Dolinar (2016), and this contains temperature and precipitation data with 1x1 km spatial resolution from 1961 onwards. Although this dataset is valuable to describe the mean climate and regimes, its monthly temporal scale is insufficient to address

65   the frequency, duration and intensity of the extremes, which clearly define the variable climate of Slovenia. Based on the

experience of previous studies creating daily datasets at the regional level, both for precipitation (Chaney et al., 2014; Hernández et al., 2016; Serrano-Notivoli et al., 2017; Yatagai et al., 2012) and temperature (Lussana et al., 2018; Serrano-Notivoli et al., 2019), a new high-resolution dataset for the Slovenian territory is suitable considering the high-density network of stations that is available. Serrano-Notivoli et al. (2017, 2019) developed a method to reconstruct and create daily gridded

70   datasets using all the available information in three stages: 1) quality control of raw data; 2) estimate of missing values and 3) prediction of new values over a gridded dataset. This methodology has been proved to work well in dense networks such as the one available for mainland Spain and its island territories (Serrano-Notivoli et al., 2017, 2019), but also for specific mountain areas with a large diversity of environments (Cuadrat et al., 2020; Martínez del Castillo et al., 2019).

We present a new high-resolution daily gridded precipitation and temperature dataset for Slovenia (SLOCLIM) with a spatial

75   resolution of 1x1 km for the period 1950 to 2018. We used all the available climatic information in the country gathered from stations in all possible environments to create a grid based on thoroughly reconstructed data series in three steps: a) a comprehensive quality control of the original data; b) the estimate of new values corresponding to the missing data; and c) the creation of the gridded dataset using the reconstructed series.

2 Input data

80   We used the meteorological records from 174 stations of the Slovenian Meteorological Agency (ARSO) regularly distributed over the territory of Slovenia (Figure 1) to create the gridded datasets. Although some data series begin before 1950, we decided to limit the research to the years 1950 to 2018, when the station network remained stable over time and space.
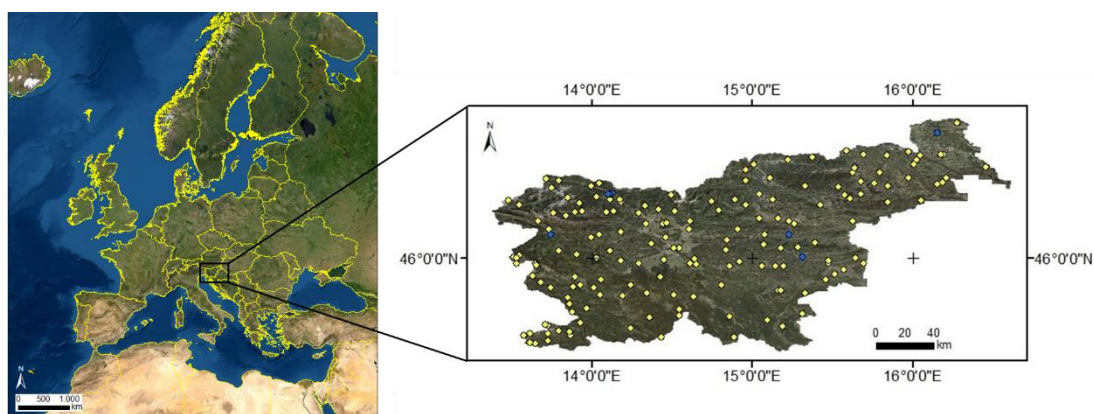
Figure 1: Location of Slovenia in Europe (left) and spatial distribution (right) of the 174 meteorological stations with temperature and precipitation information (yellow dots) and only precipitation (blue dots) (Esri, 2020; GISCO, 2020; GURS, 2020).

All the stations recorded daily precipitation, while 167 recorded both precipitation and temperature (ARSO, 2020). The altitudinal pattern showed a lower availability of data at the highest elevations, with less than five stations over 1,000 m a.s.l.

90 (Figure 2b) and corresponding with less populated and afforested areas. The average amount of annual data from stations also varied over time (Figure 2a), with a rapid increase in the first 20 years, reaching the highest availability in 1970 (107) and decreasing thereafter. The early measurements contained more precipitation than temperature data.
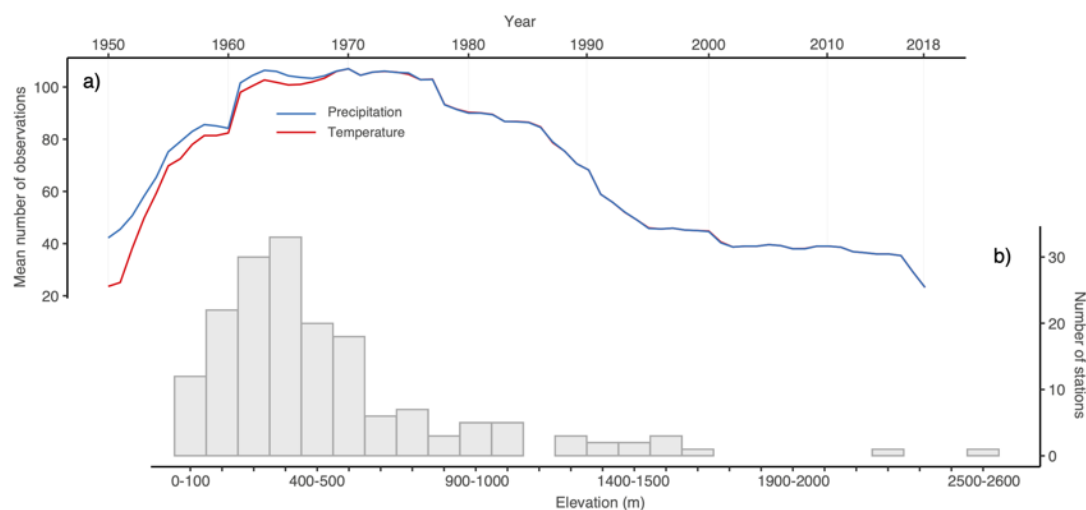
Figure 2: Original data: a) Mean annual number of observations of precipitation and temperature, and b) number of stations by
95 **elevation ranges.**

3 Methodology

A complete reconstruction of the original daily data series was performed following the three-step methodology developed in Serrano-Notivoli et al. (2017, 2019), which is based on an (1) exhaustive quality control (QC) of the raw series, (2) the estimate of new values for missing records using all the available information and local regression models, and (3) the estimates of new
100 values in a regular grid using the reconstructed series.

The QC detected and removed the most obvious erroneous data in the original dataset based on different criteria created to keep the internal and the spatial coherence of the data. For precipitation five criteria were used: (1) suspect data: the observed value was over zero and all its 10 nearest observations were zero; (2) suspect zero: the observed value was zero and all its 10 nearest observations were over zero; (3) suspect outlier: the magnitude of the observed value was 10 times higher or lower
105 than that one predicted by its 10 nearest observations; (4) suspect dry day: the observed value was zero, wet probability was over 99%, and predicted magnitude was over 5 mm; and (5) suspect wet day: the observed value was over 5 mm, dry probability was over 99%, and predicted magnitude was under 0.1 mm. For temperature we used also five criteria: (1) internal coherence; (2) removal of months containing less than three days of data; (3) removal of those days out of range considering maximum temperature (T_{MAX}) $\geq$ 50 °C or T_{MAX} $\leq$ –30 °C and minimum temperature (T_{MIN} $\geq$ 40 °C or T_{MIN} $\leq$ –35 °C; (4) removal of all
110 days in a month with a standard deviation equal to zero (suspect repeated values in the series); and (5) removal of all days in a month if the sum of the differences between T_{MAX} and T_{MIN} was equal to zero (suspect duplicated values in T_{MAX} and T_{MIN}).

These criteria helped us to evaluate, amongst others, situations of internal incoherence like abnormal values, significant deviations from other values or duplicated values. In such cases the data was removed. In order to perform the QC process by applying the above referenced criteria, reference values (RVs) for each day and the location of the original dataset were then

115 calculated using the information from the 10 nearest stations, whose selection varied according to the data availability. The RVs were estimated through the combined use of generalized linear mixed models (GLMMs) and generalized linear models (GLMs). The precipitation or temperature data was used as a dependent variable in each case, and the geographic information of each station (latitude, longitude, altitude and distance from the coast) as the independent variables. The calculated RVs were then compared with their corresponding original data to assess their quality and remove suspect values following the QC

120 criteria.

After QC, the second stage consisted of calculating new RVs based on the cleaned dataset, that were used to replace the missing values. A leave-one-out cross-validation (LOO-CV) was implemented, where RVs were calculated for all days and locations including those for which an observation exists but without using that observation, for comparison and validation purposes. This whole process produced a serially-complete dataset.

125 In the third stage, new values at each grid location of the 1x1 km gridded dataset were estimated based on the reconstructed point-based dataset, including a measure of uncertainty for each location and day, as calculated from the standard error of the models. The advantage of this methodology is that all of the available data can be used as there are no restrictions imposed due to the length or structural characteristics of the series.

Based on the gridded dataset created from the original data and with the described method, we computed four temperature and

130 two precipitation indices to show the potential applications of the grid: (1) the mean annual maximum value of daily maximum temperature, (2) the mean annual minimum value of daily minimum temperature, (3) the average annual count of days when daily minimum temperature was below 0 °C (frost days), (4) the average annual count of days when daily maximum temperature was over 25 °C (summer days), (5) maximum number of consecutive dry days per year with daily precipitation of less than 1 mm, and (6) daily precipitation intensity (annual precipitation/number of wet days). All the indices were

135 presented with their corresponding uncertainty estimates.

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4 Results

140 ### 4.1 Quality control

Precipitation:

The number of removed data varied over the years, increasing from 1950 onwards and reaching a peak in 1967, and then decreasing until the end of the period (Figure 3a). The majority of the data was removed due to the *Suspect data* and *Suspect*

zeros criteria, which usually represent estimates of low precipitation values based on nearest observations. While these criteria

145 are commonly frequent in this kind of QC (Serrano-Notivoli et al., 2017, 2019), only 1.26% of the original raw observations were removed.

Temperature:

The quality control process removed 1.28% of the original T_{MAX} data and 0.86% of the original T_{MIN} data (Figure 3b). According to the monthly criteria, 1,090 (0.06%) pieces of data for the T_{MAX} and 2,139 (0.13%) for the T_{MIN} were removed,

150 and according to the daily criteria, 20,491 (0.21%) values were removed for the T_{MAX} and 12,283 (0.73%) values for the T_{MIN}.
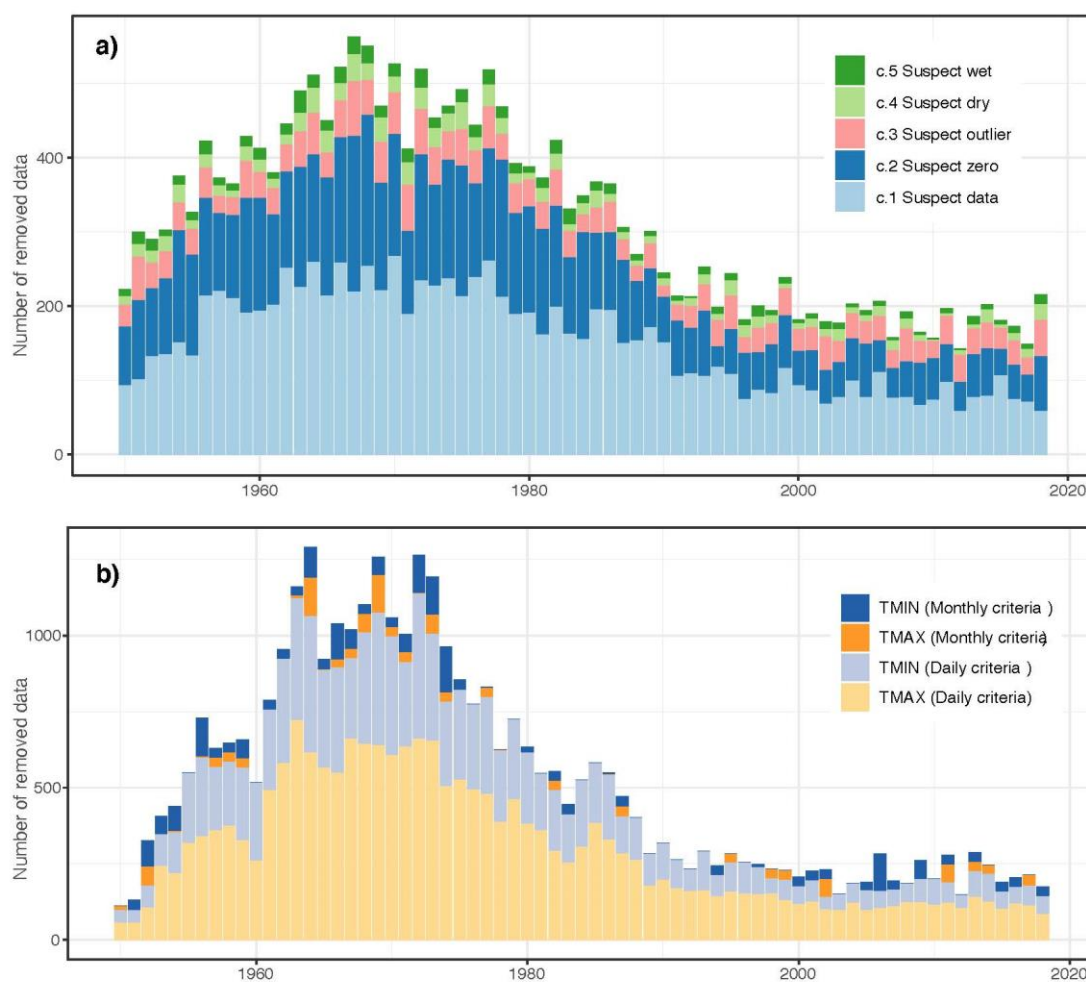
Figure 3: Number of removed precipitation (a) and temperature (b) data from the original dataset by the different criteria. The criteria are explained in more detail in Serrano-Notivoli et al. (2017, 2019).

155 **4.2 Measurements – estimation comparison**

Precipitation:

The number of observed zero precipitation days (dry days) in Slovenia was 1,006,891, and the number of estimated ones (only for corresponding days with measurements) was 1,011,337, the difference being +0.44% in favour of estimated ones. Terming the wet days as positive (observed P (precipitation) >0) and the dry days as negative (observed P=0), the true negative rate

160   (RV=0 & P=0) was over 92%, and the true positive rate (RV>0 & P>0) over 85% (Table 1). To a large extent the false negatives (RV=0 & P>0) and false positives (RV>0 & P=0) were due to the prediction of precipitation for days with low amounts of this. In events with very low amounts of precipitation, the estimate of the probability of occurrence was likely to be dry, despite the fact that the station could register a minimum quantity of rain (usually under 1 or 2 mm). This causes a small difference in amounts, and is becoming more distinct in the dry/wet accuracy assessment.

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Table 1: Accuracy of the wet/dry day estimates: percent of observed dry (P (precipitation) = 0) and wet (P > 0) days, and percent of predicted dry (RV = 0) and wet (RV > 0) days on observed dry and wet days. (Pred – predicted, Obs – observed).

	Obs = 0	Obs > 0
Pred = 0	92.55	14.13
Pred > 0	7.45	85.87

The highest Pearson correlations (> 0.94) were found in the comparison by days between the original dataset and corresponding

170   estimates, rather than in the comparison by stations (Figure 4). This indicates the complexity of the geographical variability (elevation, orientation, landscape configuration, etc.) between stations. The highest correlation between the measurements and estimates by stations was found for medians without zeros (only considering days with P>0 in measurements and estimations). The comparison showed the lowest correlation in the 95th percentiles of wet days when comparing by both stations and days.

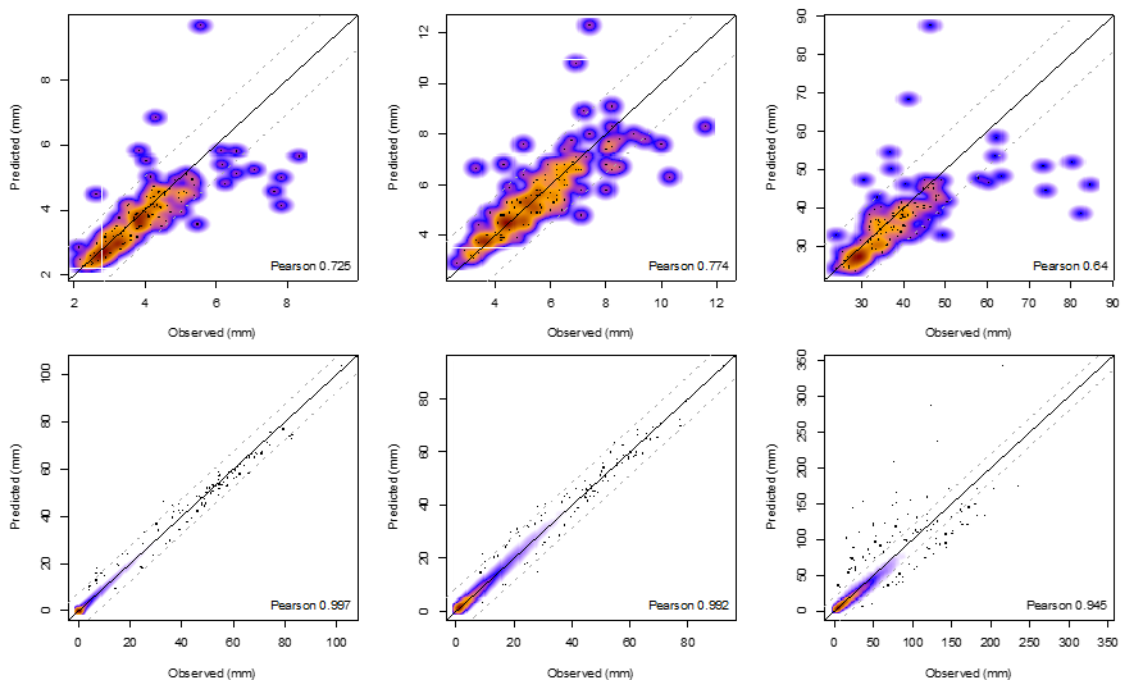



Figure 4: Scatterplots and Pearson correlation coefficients between measurements and estimations of daily precipitation by stations (upper row) and days (lower row) of means (left column), medians without zeros (central column) and 95th percentiles (right column).

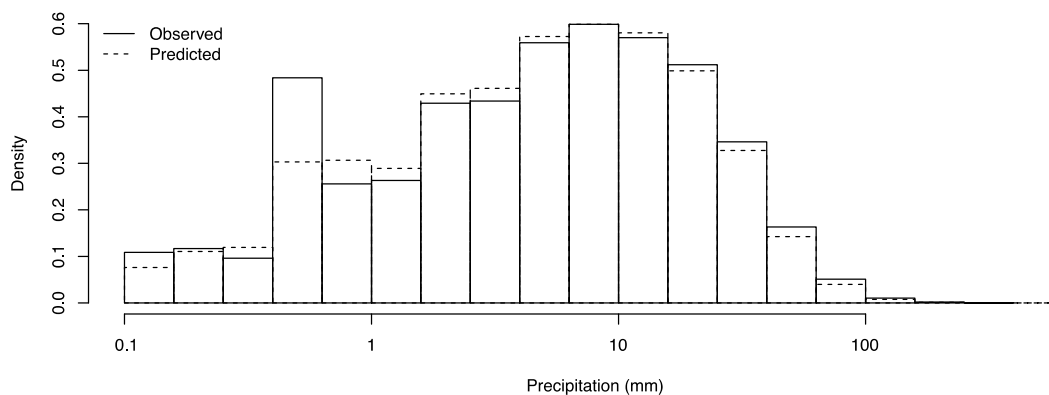



Figure 5: Histograms of observed and predicted daily precipitation frequency.

The histogram of estimated and observed precipitation showed a good agreement (Figure 5). However, values below 1 mm were slightly underestimated in some cases. The agreement between the histograms was very high above 1 mm.

The fit of the estimated values by altitudinal ranges showed a slight overestimation in the ratio of means (RM) between 700 and 900 m a.s.l. (Table 2). Overestimation was higher above 1100 m a.s.l. with the exception of 1500-2000 m a.s.l., where the

185 estimation was lower than the measurements. The ratio of the standard deviation (RSD) also showed differences between estimated and observed values, especially between 1100 and 1300 m a.s.l. The mean absolute error (MAE) increased with altitude, which can be attributed to the decreasing number of available stations with altitude.

190 **Table 2: The leave-one-out cross-validation (LOO-CV) statistics showing the goodness of fit between measurements and estimations of daily precipitation separated by altitudes (m a.s.l.). N: number of stations; %OBS: percentage of observed precipitation; %PRE: percentage of predicted precipitation; MAE: mean absolute error; ME: mean error; RM: ratio of means; RSD: ratio of standard deviations. The results were rounded to two decimal places.**

Altitude	N	%OBS	%PRE	MAE	ME	RM	RSD
0-300	48	33.30	34.20	1.33	-0.01	1.00	0.97
300-500	43	28.70	27.10	1.41	-0.31	0.91	0.86
500-700	18	15.50	14.50	1.61	-0.45	0.90	0.87
700-900	8	7.90	8.80	2.38	0.39	1.09	1.16
900-1100	8	8.30	7.30	2.72	-0.84	0.85	0.81
1100-1300	2	0.80	1.20	3.16	1.36	1.37	1.38
1300-1500	2	1.80	2.10	3.17	0.59	1.13	1.28
1500-2000	3	3.60	2.70	3.72	-1.78	0.74	0.67
>2000	1	2.10	3.70	6.05	4.11	1.74	1.95

195 **Table 3: The leave-one-out cross-validation (LOO-CV) statistics for precipitation, showing the goodness of fit between measurements and estimates of monthly aggregates. MAE: mean absolute error; ME: mean error; RM: ratio of means; RSD: ratio of standard deviations. The results were rounded to two decimal places.**

	MAE	ME	RM	RSD	Pearson	Range
Jan	1.93	-0.32	0.96	0.95	0.82	0 to 78.3
Feb	1.74	-0.29	0.96	0.95	0.86	0 to 75.5
Mar	1.95	-0.26	0.97	0.96	0.84	0 to 65.8
Apr	2.35	-0.35	0.97	0.99	0.78	0.2 to 74.1
May	2.62	-0.34	0.97	0.99	0.78	0.2 to 49.8
Jun	3.08	-0.58	0.96	0.97	0.76	0.7 to 51.9
Jul	2.95	-0.61	0.95	0.98	0.79	0.3 to 49.2
Aug	2.86	-0.58	0.96	0.96	0.82	0 to 69.7
Sep	2.94	-0.44	0.97	0.96	0.85	0.2 to 116
Oct	3.04	-0.38	0.97	0.97	0.85	0 to 97.2

| Nov | 3.21 | -0.38 | 0.97 | 0.97 | 0.80 | 0 to 115.5 |
| Dec | 2.50 | -0.35 | 0.97 | 0.94 | 0.82 | 0 to 81.4 |

The ratio of means (RM) indicated that the monthly aggregates of daily precipitation in all months were slightly underestimated. The ratio of standard deviation (RSD) was very close to 1 through all months (Table 3). In December, however,

200 the RSD was below 0.95, indicating a bias in the variance estimate. In the other months there are no substantial biases in the variance estimation. Overall, all monthly statistics are relatively homogeneous over the year due to the evenly distributed precipitation over the year.

Temperature:

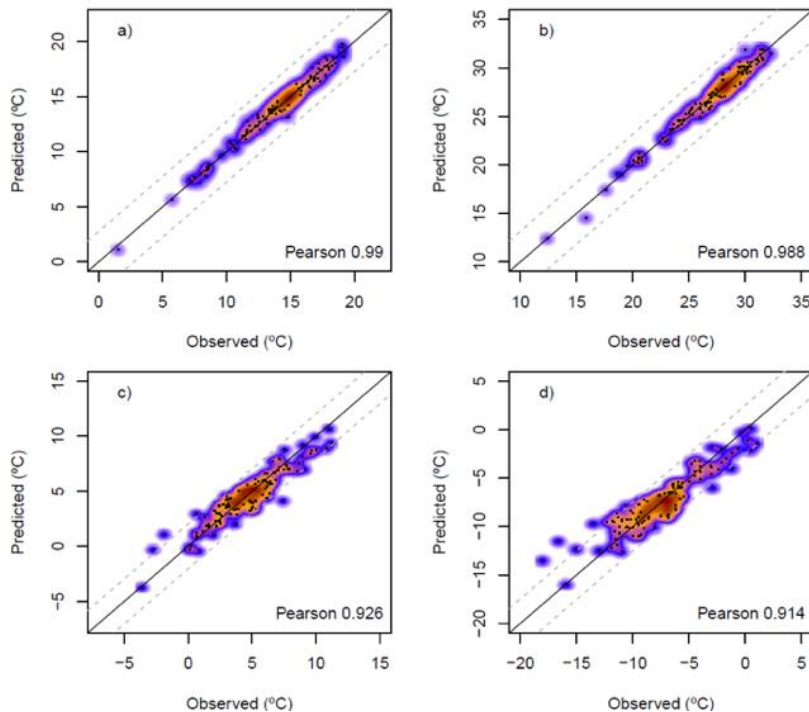
205 The results showed very high correlations between the estimated values and measurements both for T_{MAX} and T_{MIN} when considering daily means, as the Pearson correlation coefficients were 0.99 and 0.93, respectively (Figure 6). Similarly, good agreements were obtained when considering the extremes (5th percentile of T_{MIN} and 95th percentile of T_{MAX}) with Pearson correlations of 0.99 and 0.91, respectively.

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Figure 6: Comparison between measurements and estimates, by stations, of the mean maximum temperatures (a) and their 95th percentiles (b), and comparison of the mean minimum temperatures (c) and their 5th percentiles (d). Dashed lines represent ±1 standard deviation of the data.
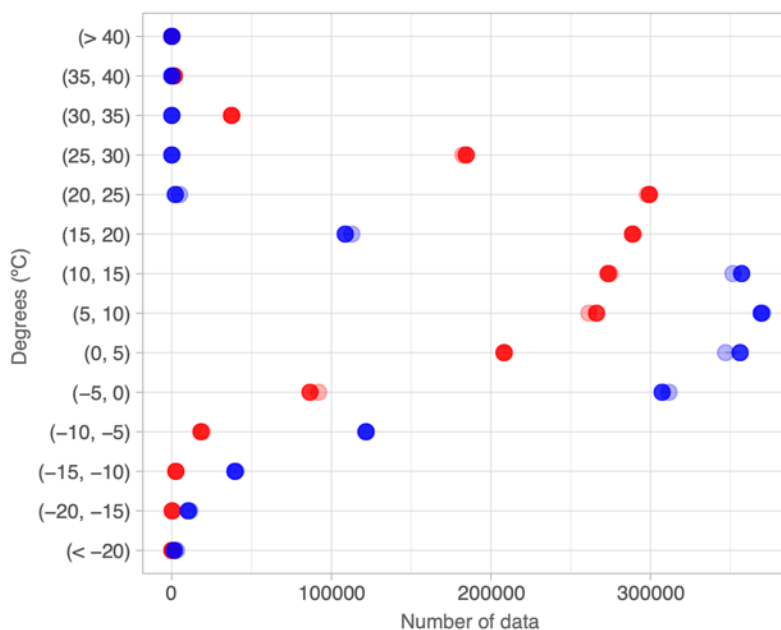
Figure 7: Comparison of frequencies by categories between observed (solid) and predicted (transparent) maximum (red) and minimum (blue) temperatures.

The frequency of the observed temperatures and their estimates (Figure 7) showed very good general agreement. There was a slight underestimation of T_{MIN} (in the temperature intervals 0 to 5 °C and 10 to 15 °C) and T_{MAX} (interval 5 to 10 °C). Overestimation was found for T_{MAX} and T_{MIN} for the interval -5 to 0 °C.

Table 4: The leave-one-out cross-validation (LOO-CV) statistics showing the goodness of fit between measurements and estimates of daily maximum and minimum temperature separated by altitudes (m a.s.l.). N: number of stations; OBSm: mean observed maximum and minimum temperature (maximum and minimum temperature in °C); PREDm: mean predicted temperature (°C); MAE: mean absolute error (°C); ME: mean error (°C); RM: ratio of means; RSD: ratio of standard deviations. The results were rounded to two decimal places.

	Altitude (m)	0–100	100–300	300–500	500–700	700–900	900–1100	1100–1300	1300–1500	1500–2000	> 2000
T_{MAX}	OBSm	18.30	15.80	14.70	13.50	12.00	11.30	9.30	8.70	7.30	7.30
	PREDm	18.50	15.80	14.60	13.70	12.20	11.50	9.40	8.70	7.50	7.10
	MAE	0.56	0.35	0.34	0.45	0.44	0.36	0.36	0.50	0.40	0.46
	ME	0.50	-0.02	-0.01	0.14	0.04	0.06	0.13	0.03	-0.01	-0.20
	RM	1.03	0.99	0.99	1.01	0.99	1.01	1.02	0.99	0.99	0.96
	RSD	1.03	1.01	0.99	0.98	1.01	0.99	0.99	1.02	0.99	0.76
	N	12	61	59	35	11	10	3	4	4	2
T_{MIN}	OBSm	9.60	5.50	4.60	4.30	3.50	3.30	3.10	0.80	0.90	1.80
	PREDm	8.80	5.50	4.70	3.90	3.80	2.60	3.00	0.99	0.70	1.70

MAE	1.04	0.57	0.69	0.84	1.06	0.79	0.59	0.60	0.74	0.32
ME	-0.18	-0.02	0.10	-0.30	0.25	-0.43	-0.13	0.27	-0.63	-0.13
RM	0.98	0.99	1.02	0.92	1.07	0.81	0.96	1.19	0.62	0.99
RSD	0.99	0.99	0.99	0.99	0.99	0.98	0.97	0.99	1.01	1.01
N	12.00	61.00	59.00	35.00	11.00	10.00	3.00	4.00	4.00	2.00

Most of the stations in Slovenia are located at lower altitudes, 60% of them between 100 and 500 m a.s.l., which leads to a slight underestimation of T_{MIN} values at 900 – 1100 and 1500-2000 m a.s.l. due to having less data. Also, an overestimation of T_{MIN} was observed at the 1300 –1500 m a.s.l. range, but the similarity between the observed and estimated temperatures was generally higher for T_{MAX} (Table 4).

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Table 5: The leave-one-out cross-validation (LOO-CV) statistics showing the goodness of fit between maximum and minimum temperature measurements and estimates of monthly aggregates. MAE: mean absolute error (°C); ME: mean error (°C); RM: ratio of means; RSD: ratio of standard deviations; Pearson: Pearson correlation coefficient. The results were rounded to two decimal places.

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		Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov
T_{MAX}	MAE	0.48	0.42	0.33	0.31	0.31	0.31	0.33	0.32	0.33	0.37	0.4
	ME	0.05	0.07	0.01	0.00	0.03	0.02	0.02	0.03	0.03	0.06	0.03
	RM	1.01	1.01	0.99	0.99	0.99	0.99	0.99	0.99	0.99	0.99	0.99
	RSD	0.97	0.97	0.98	0.99	0.99	0.97	0.98	0.97	0.99	0.97	0.97
	Pearson	0.98	0.99	0.99	0.99	0.99	0.98	0.98	0.99	0.99	0.98	0.98
T_{MIN}	MAE	0.78	0.74	0.71	0.67	0.75	0.66	0.69	0.67	0.7	0.68	0.59
	ME	-0.05	-0.07	0.03	-0.02	0.01	-0.04	-0.04	-0.07	-0.07	-0.07	0.00
	RM	0.99	0.97	0.71	0.97	0.99	0.99	0.99	0.99	0.99	0.99	0.89
	RSD	0.99	0.99	0.96	0.93	0.93	0.93	0.94	0.93	0.94	0.96	0.95
	Pearson	0.98	0.98	0.97	0.95	0.93	0.94	0.93	0.93	0.95	0.96	0.97

The ratio of means (RM) showed that the differences in the means between observed and estimated temperatures were lower for T_{MAX}, with all the values very close to 1 (Table 5). T_{MIN} showed, overall, a higher underestimation, especially in March. Moreover, the ratio of standard deviation (RSD) was higher in T_{MAX}, while T_{MIN} showed values under 0.95 from April to

240 September, indicating a bias in variance estimation in these months. ME values were close to 0 in all cases, and Pearson correlations were higher than 0.98 in T_{MAX} in all months, ranging from 0.93 to 0.98 in T_{MIN}.

4.3 Spatial distribution and uncertainty in daily precipitation

Mean precipitation on wet days (SDII) showed that the highest values occurred in the central to northwestern part of the country, and also at some individual locations on the south and north, with amounts up to 27 mm (Figure 8a). The lowest
245 amounts were in the northeastern part of Slovenia (less than 10 mm). The average maximum length of dry spells (CDD) reached the highest values in the northeastern part of the country, with 70 days, while this number decreased in the southeast and northwest until the values were lower than 35 days (Figure 8b). The uncertainty of SDII was very low (<10%) all over the country (Figure 9a), while the uncertainty of CDD averaged between 16 and 20% in most of it, with higher values (21–25%) in the northeastern part (Figure 9b).

250 The highest values of the mean annual absolute daily maximum temperature (TX_x) were found in the southwest and east of Slovenia, with temperatures up to 36 °C (Figure 10a), while values below 20 °C were more frequent at higher elevations. As expected, the mean annual absolute daily minimum temperature (TN_n) showed a similar pattern to TX_x (Figure 10b), with the lowest values in the north and northwestern part of Slovenia (-22 °C) and much higher values in the southwest, where the influence of the sea affects temperatures. However, in the eastern half of the territory there was a lower spatial variability
255 compared to that seen for TX_x. The mean annual number of summer days (SU) was calculated from the annual count of days when daily T_{MAX} was over 25 °C. Similar to the TX_X, the highest values were found in the southwestern and also eastern part of the country (Figure 10c). In the mountain areas, however, only a small number of summer days were found. The mean annual number of frost days (FD) was calculated from the annual count of days when the minimum daily temperature was below 0 °C. The number of frost days varied from 36 in the coastal area to 240 in the highest elevations, where orography had
260 a significant influence (Figure 10d).

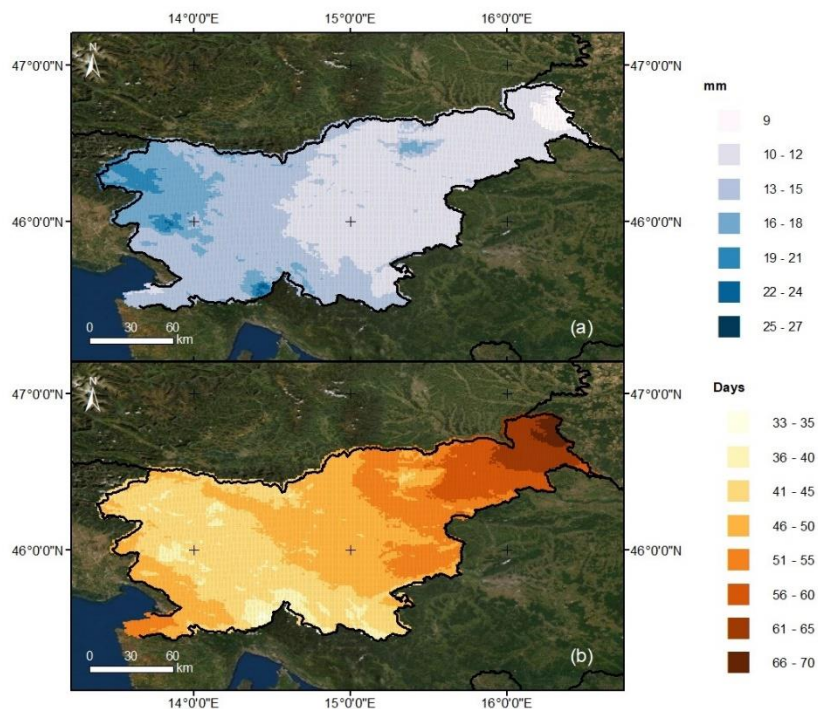
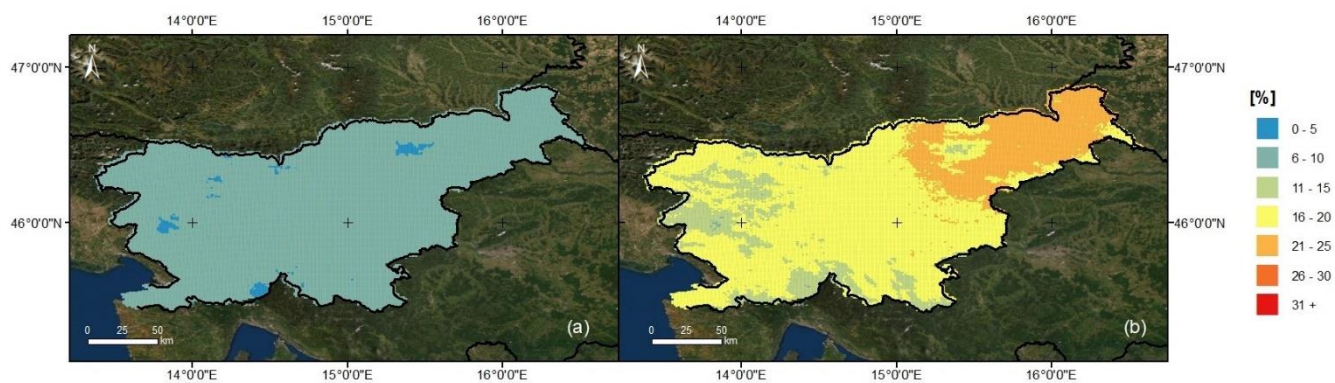
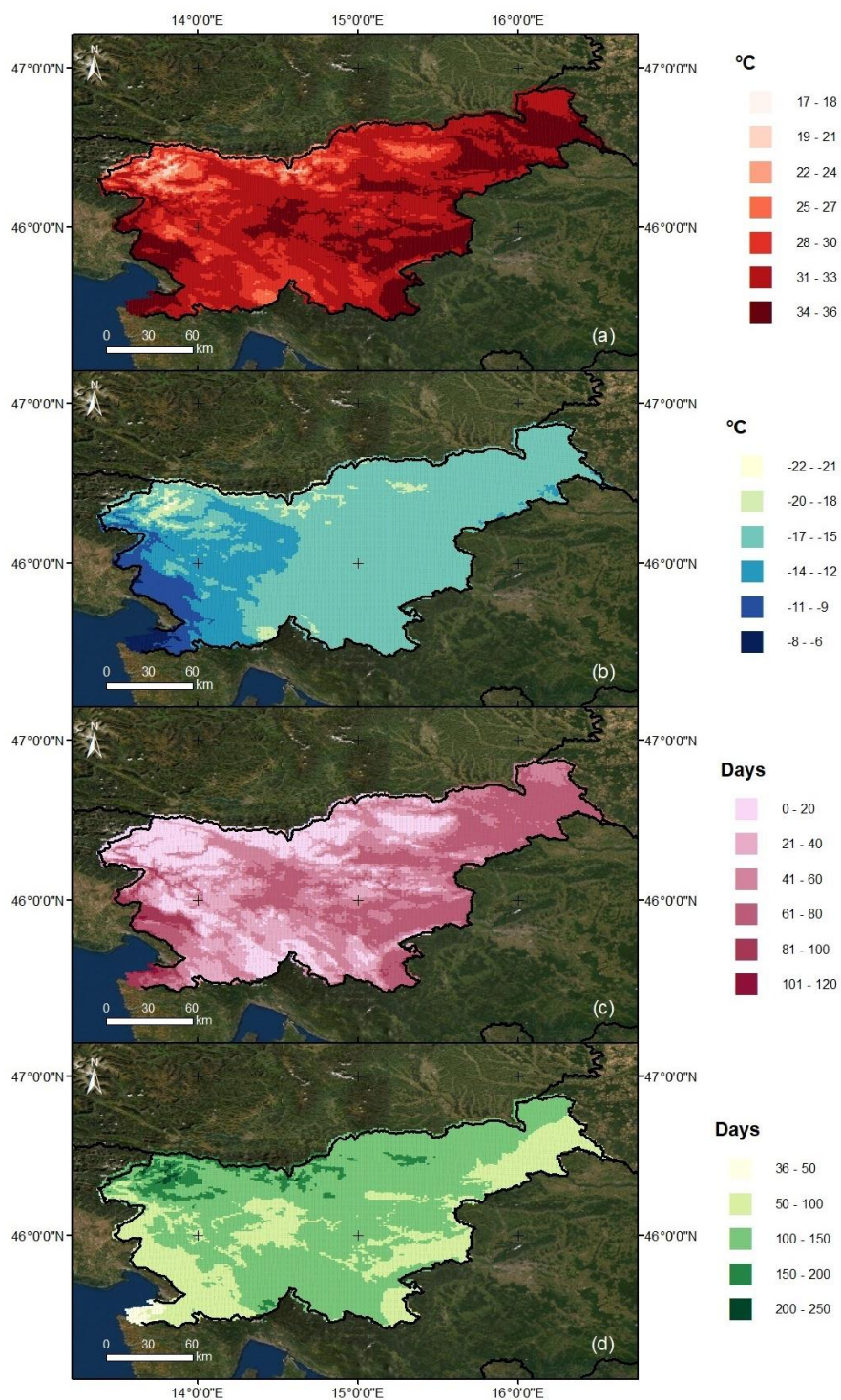
The highest uncertainty values of TX_x and SU were found in the northwestern part of the country (Figure 11a, c), while the highest values of uncertainty of TN_n were found in the southwestern part (Figure 11b). Uncertainty was very low for the number of frost days (Figure 11d).

265

Figure 8: Daily precipitation indices: (a) SDII - daily precipitation intensity, (b) CDD - maximum number of consecutive dry days (Esri, 2020; GISCO, 2020).

270   **Figure 9: Uncertainty of SDII: a) daily precipitation intensity and (b) CDD - maximum number of consecutive dry days (Esri, 2020; GISCO, 2020).**

275 **Figure 10: Mean annual (1950-2018) values for (a) TXx - absolute daily maximum temperature, (b) TNn – absolute daily minimum temperature, (c) SU - number of summer days, (d) FD - number of frost days (Esri, 2020; GISCO, 2020).**

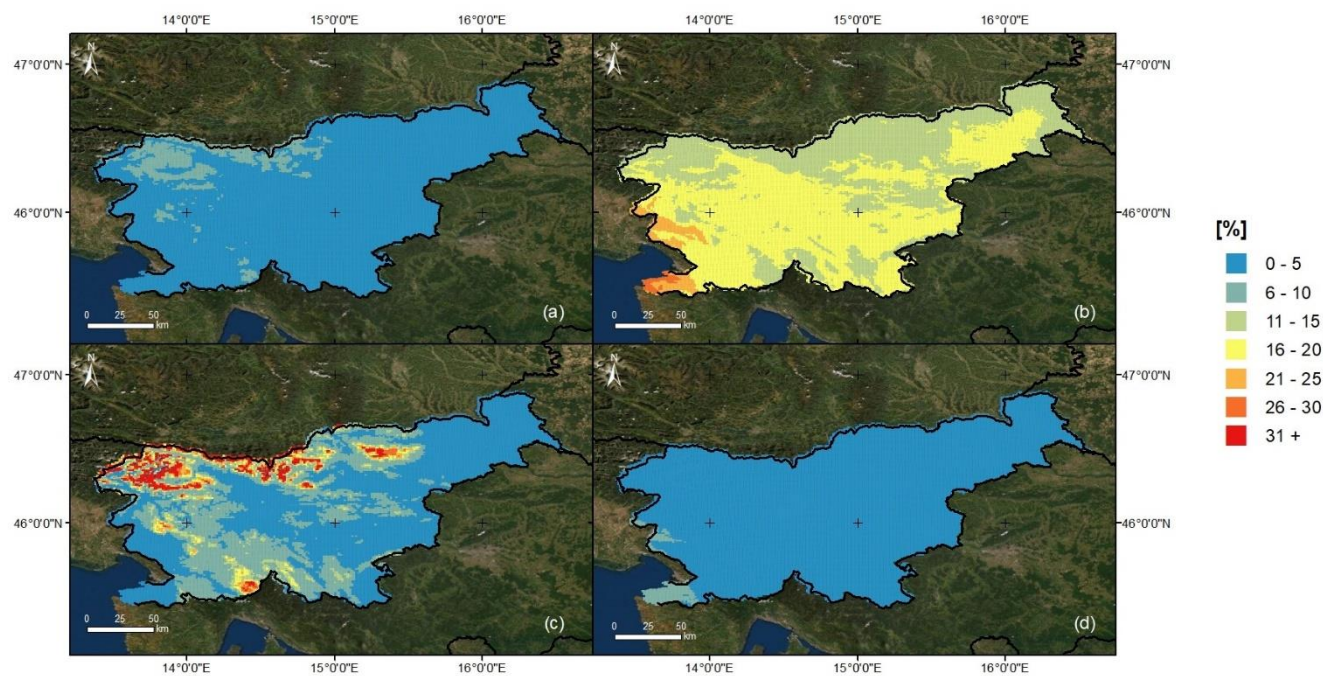



Figure 11: Uncertainty of a) TXx - absolute daily maximum temperature, b) TNn - absolute daily minimum temperature, c) SU - number of summer days and d) FD - number of frost days (Esri, 2020; GISCO, 2020).

280

5 Discussion

SLOCLIM is the first climatic reconstruction for Slovenia at a high spatial resolution, providing daily data of T_{MAX}, T_{MIN} and the amount of precipitation from 1950 to 2018. This new daily gridded dataset contributes significantly to the climate description of the country, and it is expected to be of high value for multiple potential applications, especially in areas where observed climate records are not available, such as agricultural land, forests and mountains.

There were no climatic reconstructions available for Slovenia until the work of Dolinar (2016), who established a grid on a 1 km scale for the monthly temperature and precipitation from 1961 onwards, using explanatory variables such as longitude, latitude, altitude and their second degree polynomial terms to improve the reconstruction. Besides this, there have been other activities aimed at the homogenization of climatic data, such as the project "Climate Change and Variability in Slovenia" (Vertačnik et al., 2015), which was based on homogenized monthly data with the goal of obtaining high quality meteorological data from 1961 onwards.

In order to fill the gap, the SLOCLIM database was created with a methodology that is suitable for a country like Slovenia with large climatic and orographic differences, as it provides a reliable independent assessment of climate parameters,

considering the local character of precipitation and temperature at each particular location. The period 1950 to 2018 represents
the longest possible time-span for data reconstruction at the given resolution, which is mainly determined by the number,
location and quality of meteorological stations. The method makes use of all available information and includes a
comprehensive iterative quality control process that checks the spatial and temporal consistency of the data until no suspect
values are detected.

The presented results show very high correlations between the estimated values and the measurements for both temperatures
and precipitation, indicating the ability to reproduce realistic climatic situations. The comparison by days showed higher
Pearson correlations between the original precipitation dataset and the corresponding estimates, than the comparison by
stations. The reason for this is the complexity of the high geographical variability between stations, which could, to a certain
extent, also influence the underestimation of minimum temperatures at higher altitudes. The dataset used in Dolinar (2016)
showed slightly higher correlation coefficients between predicted and measured values than found in our study, but it was
constructed on a monthly basis using a completely different approach. Furthermore, Dolinar (2016) obtained lower correlations
between predicted and measured values for precipitation than for temperature, which is consistent with our results and is
attributed to the generally higher precipitation variability both in space and time.

The agreement between the original dataset and the corresponding estimates in our study was high due to the good quality of
the original data, mainly because ARSO regularly undertakes a validation control of the climatological data collected by
meteorological stations, resulting in consolidated data (Bertalanič et al., 2006). Although the spatial coverage with
meteorological stations decreases over time, it is still relatively high compared to other countries.

In Slovenia, precipitation is highest in the northwest (in the Julian Alps) and southwest in the Dinaric Mountains, and lowest
in the northeast (ARSO, 2016b), as confirmed also by our reconstruction. The highest uncertainties in precipitation were found
in the northeast, in the area with the lowest amount and intensity of precipitation. Therefore, the uncertainty is higher because
it is more difficult to make solid predictions in areas where precipitation is very low and the number of consecutive dry days
is higher. The highest temperatures are reached in the southwestern part of the country, and the lowest in the northwest (in the
Alpine region), as shown by ARSO (2016a) and confirmed by our reconstruction. The presented indices of mean annual
absolute daily maximum and minimum temperatures are in agreement with the observations of ARSO (2016a). The highest
uncertainty levels for the high temperature indices (TXx, SU) were found for the northwest, and for the lowest temperatures
(TNn, FD) in the southwest, possibly due to the high orographic effect and the distance to the sea. In the southwest, lower
temperatures occur less frequently then in other parts of the country, which possibly explains why the uncertainty is greater.
In this regard, the uncertainty of the number of summer days (SU) was particularly high in the areas at the highest altitudes.
The geographical distribution of the temperature indices is consistent with the climatic regions defined by Kozjek et al. (2017).

The SLOCLIM database opens up many possibilities for applications in areas which need data with high spatial and temporal
resolution. It has the potential to address a number of relevant climate related topics, including recognition of climatic trends
and extremes as well as their mitigation and planning of prevention measures. SLOCLIM is expected to support ongoing and
future research activities which are focused on basic climatic questions in Slovenia and surrounding regions, such as like trend

analyses and projections (Dolinar et al., 2018; de Luis et al., 2014; Milošević et al., 2017), as well as reconstruction of past climate and extreme events based on tree-ring parameters (Cook et al., 2015; Čufar et al., 2008; Hafner et al., 2014). Research

330 where climate data are needed also includes that focused on the effects of climate change on the well-being and health of population (Ciuha et al., 2019; Pogačar et al., 2019, 2020), as well as agroclimatic shifts and the food supply (Ceglar et al., 2012, 2019a, 2019b). In Slovenia, a densely afforested country, climate affects the survival and productivity of forest trees due to shifts in leaf phenology (Čufar et al., 2012, 2015), and the production of crucial tree tissues, like wood and phloem (Delpierre et al., 2019; Martinez del Castillo et al., 2018; Prislan et al., 2013). While forest trees can be damaged due to extreme events

335 like ice storms (Decuyper et al., 2020; Klopčič et al., 2020), the effects of climate differ along various gradients (Čater and Levanič, 2019; Jevšenak et al., 2020) and at the microsite level (Diaci et al., 2020; Kermavnar et al., 2019). In most cases the use of precise daily climatic parameters can help to improve recognition of climate-growth relationships (Jevšenak, 2019; Jevšenak and Levanič, 2018).

6 Data availability

340 The SLOCLIM dataset is freely available in the web repository Zenodo. It can be accessed through https://doi.org/10.5281/zenodo.4108543 and cited as Škrk et al. (2020). The data are arranged in six files (daily maximum and minimum temperature, daily amount of precipitation and their uncertainties).

7 Conclusions

The result of the presented study is a high-resolution daily gridded precipitation and temperature dataset for Slovenia with a

345 spatial resolution of 1x1 km for the period 1950 to 2018 (SLOCLIM). Advantages of SLOCLIM compared to other existing datasets in Slovenia are the use of all available data and the provision of reconstructed daily data over a dense grid. It delivers the daily amount of precipitation as well as maximum and minimum temperature. The available measurement data with regard to latitude, longitude, altitude, and distance from the coast were used to calculate the dataset. All the data were quality controlled by using cross-validation. Missing values were calculated using the 10 nearest observations. All results showed high

350 correlation between estimated and observed values, and six climatic indices were also calculated: precipitation intensity, length of dry period, mean annual absolute maximum and minimum temperatures, mean annual number of frost days and mean annual number of summer days. The indices correlated well with altitudes and distance from the coast.

SLOCLIM represents a novel publicly available climatic database for Slovenia with multiple potential applications for recognition and mitigation of climate related events directly affecting the environment, well-being and health of the population,

355 agriculture and food-supply, as well as forestry. High resolution gridded daily data are expected to facilitate research activities in numerous scientific disciplines studying climatic trends and projections, reconstruction of past climate and prediction of

future climate, environment, human and animal populations as well as agricultural and forest systems. The data can help to identify extreme climatic events and can be also used for prediction, adaptation and mitigation measures.

Author contributions

360   RSN, MdL, KC and NS designed the research, RSN and MdL developed the methodological approach, LKB, ZC and KC organized the original data, MdL, RSN and NS made the reconstruction of the climatic data and the developed the gridded dataset. RSN and NS made the validation processes and performed the analysis and presentation of the results. NS and RSN designed and wrote the manuscript with contribution from all co-authors.

Competing interest

365   The authors declare that they have no conflicts of interest.

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Review statement

375   **References**

ARSO: Letna povprečna temperatura zraka (1981-2010), [online] Available from: http://meteo.arso.gov.si/uploads/probase/www/climate/image/sl/by_variable/temperature/annual-mean-air-temperature_81-10.png (Accessed 11 September 2020a), 2016.

ARSO: Povprečna letna višina merjenih padavin (1981-2010), [online] Available from:
380   http://meteo.arso.gov.si/uploads/probase/www/climate/image/sl/by_variable/precipitation/mean-annual-measured-

precipitation_81-10.png (Accessed 11 September 2020b), 2016.

ARSO: Archival data on meteorological conditions in Slovenia, Slov. Environ. Agency. Minist. Environ. Spat. Planning, Ljubljana, Slov. [online] Available from: http://meteo.arso.gov.si/met/sl/archive/ (Accessed 1 March 2020), 2020.

Bertalanič, R., Dolinar, M., Nadbath, M., Ovsenik-Jeglič, T., Vertačnik, G. and Vičar, Z.: Climate of Slovenia 1971–2000, edited by M. Dolinar, p. 20, The Environmental Agency of the Republic of Slovenia, Ljubljana, Slovenia. [online] Available   from: https://meteo.arso.gov.si/uploads/probase/www/climate/text/en/publications/climate_of_slovenia_71_00.pdf, 2006.

Čater, M. and Levanič, T.: Beech and silver fir's response along the Balkan's latitudinal gradient, Sci. Rep., 9(1), 1–14, doi:10.1038/s41598-019-52670-z, 2019.

Ceglar, A., Medved Cvikl, B., Moran Tejeda, E., Vicente Serrano, S. M. and Kajfež Bogataj, L.: Assessment of multi-scale drought datasets to quantify drought severity and impacts in agriculture : a case study for Slovenia, Int. J. Spat. Data Infrastruct. Res., 7, 464–487, doi:10.2902/1725-0463.2012.07.art21, 2012.

Ceglar, A., Van der Wijngaart, R., De Wit, A., Lecerf, R., Boogaard, H., Seguini, L., Van den Berg, M., Toreti, A., Zampieri, M. and Fumagalli, D.: Improving WOFOST model to simulate winter wheat phenology in Europe: Evaluation and effects on yield, Agric. Syst., 168, 168–180, 2019a.

Ceglar, A., Zampieri, M., Toreti, A. and Dentener, F.: Observed Northward Migration of Agro-Climate Zones in Europe Will Further Accelerate Under Climate Change, Earth's Futur., 7(9), 1088–1101, doi:10.1029/2019EF001178, 2019b.

Cegnar, T.: Ekstremni vremenski dogodki, ARSO [online] Available from: http://kazalci.arso.gov.si/sl/content/ekstremni-vremenski-dogodki-4 (Accessed 12 September 2020), 2016.

Chaney, N. W., Sheffield, J., Villarini, G. and Wood, E. F.: Development of a high-resolution gridded daily meteorological dataset over sub-Saharan Africa: Spatial analysis of trends in climate extremes, J. Clim., 27(15), 5815–5835, doi:10.1175/JCLI-D-13-00423.1, 2014.

Ciuha, U., Pogačar, T., Bogataj, L. K., Gliha, M., Nybo, L., Flouris, A. D. and Mekjavic, I. B.: Interaction between indoor occupational heat stress and environmental temperature elevations during heat waves, Weather. Clim. Soc., 11(4), 755–762, doi:10.1175/WCAS-D-19-0024.1, 2019.

Cook, E. R., Seager, R., Kushnir, Y., Briffa, K. R., Büntgen, U., Frank, D., Krusic, P. J., Tegel, W., Schrier, G. Vander, Andreu-Hayles, L., Baillie, M., Baittinger, C., Bleicher, N., Bonde, N., Brown, D., Carrer, M., Cooper, R., Čufar, K., DIttmar, C., Esper, J., Griggs, C., Gunnarson, B., Günther, B., Gutierrez, E., Haneca, K., Helama, S., Herzig, F., Heussner, K. U., Hofmann, J., Janda, P., Kontic, R., Köse, N., Kyncl, T., Levanič, T., Linderholm, H., Manning, S., Melvin, T. M., Miles, D., Neuwirth, B., Nicolussi, K., Nola, P., Panayotov, M., Popa, I., Rothe, A., Seftigen, K., Seim, A., Svarva, H., Svoboda, M., Thun, T., Timonen, M., Touchan, R., Trotsiuk, V., Trouet, V., Walder, F., Wany, T., Wilson, R. and Zang, C.: Old World megadroughts and pluvials during the Common Era, Sci. Adv., 1(10), 1–10, doi:10.1126/sciadv.1500561, 2015.

Cuadrat, J. M., Serrano-Notivoli, R., Llabrés, A., Prohom, M., Saz, M. Á., Tejedor, E., Cunillera, J., Trapero, L., Pons, M.,

415   Copons, R., Lopez-Moreno, J. I., Luna, M. Y., Rodriguez, E., Gascoin, S. and Soubeyroux, J. M.: Creación de un sistema de indicadores para la caracterización del clima de los Pirineos, in Desafíos y oportunidades de un mundo en transición: Una interpretación desde la Geografía, edited by J. Farinos, J. Escribano, M. P. Penarrubia, J. Serrano, and S. Asins, p. 53, Universitat de València, Valencia., 2020.

Čufar, K., De Luis, M., Eckstein, D. and Kajfež-Bogataj, L.: Reconstructing dry and wet summers in SE Slovenia from oak
420   tree-ring series, Int. J. Biometeorol., 52(7), 607–615, doi:10.1007/s00484-008-0153-8, 2008.

Čufar, K., de Luis, M., Saz, M. A., Črepinšek, Z. and Kajfež-Bogataj, L.: Temporal shifts in leaf phenology of beech (Fagus sylvatica) depend on elevation, Trees - Struct. Funct., 26(4), 1091–1100, doi:10.1007/s00468-012-0686-7, 2012.

Čufar, K., De Luis, M., Prislan, P., Gričar, J., Črepinšek, Z., Merela, M. and Kajfež-Bogataj, L.: Do variations in leaf phenology affect radial growth variations in Fagus sylvatica?, Int. J. Biometeorol., 59(8), 1127–1132, doi:10.1007/s00484-014-
425   0896-3, 2015.

Decuyper, M., Chávez, R. O., Čufar, K., Estay, S. A., Clevers, J. G. P. W., Prislan, P., Gričar, J., Črepinšek, Z., Merela, M., de Luis, M., Notivoli, R. S., del Castillo, E. M., Rozendaal, D. M. A., Bongers, F., Herold, M. and Sass-Klaassen, U.: Spatio-temporal assessment of beech growth in relation to climate extremes in Slovenia – An integrated approach using remote sensing and tree-ring data, Agric. For. Meteorol., 287, 107925, doi:10.1016/j.agrformet.2020.107925, 2020.

430   Delpierre, N., Lireux, S., Hartig, F., Camarero, J. J., Cheaib, A., Čufar, K., Cuny, H., Deslauriers, A., Fonti, P., Gričar, J., Huang, J. G., Krause, C., Liu, G., de Luis, M., Mäkinen, H., del Castillo, E. M., Morin, H., Nöjd, P., Oberhuber, W., Prislan, P., Rossi, S., Saderi, S. M., Treml, V., Vavrick, H. and Rathgeber, C. B. K.: Chilling and forcing temperatures interact to predict the onset of wood formation in Northern Hemisphere conifers, Glob. Chang. Biol., 25(3), 1089–1105, doi:10.1111/gcb.14539, 2019.

435   Diaci, J., Rozman, J. and Rozman, A.: Regeneration gap and microsite niche partitioning in a high alpine forest: Are Norway spruce seedlings more drought-tolerant than beech seedlings?, For. Ecol. Manage., 455(November 2019), 117688, doi:10.1016/j.foreco.2019.117688, 2020.

Dolinar, M.: Monthly gridded datasets for temperature and precipitation over Slovenia, Proc. GeoMLA–Geostatistics Mach. Learn. Belgrade, Serbia, 2016.

440   Dolinar, M., Gregorič, G., Honzak, L., Sušnik, A., Vlahović, Ž. and Žust, A.: Ocena podnebnih sprememb v Sloveniji do konca 21. stoletja: povzetek dejavnikov okolja z vplivom na kmetijstvo in gozdarstvo, Ministrstvo za okolje in prostor, Agencija Republike Slovenije za okolje, Ljubljana. [online] Available from: https://meteo.arso.gov.si/uploads/probase/www/climate/text/sl/publications/povzetek-podnebnih-sprememb-agro.pdf, 2018.

445   Esri: Imagery [basemap], World Imag. [online] Available from: https://www.arcgis.com/home/item.html?id=10df2279f9684e4a9f6a7f08febac2a9, 2020.

Giorgi, F.: Climate change hot-spots, Geophys. Res. Lett., 33(8), 1–4, doi:10.1029/2006GL025734, 2006.

GISCO: Countries 2020, Geogr. Inf. Syst. Comm. [online] Available from:

https://ec.europa.eu/eurostat/web/gisco/geodata/reference-data/administrative-units-statistical-units/countries
450   (Accessed 15 November 2020), 2020.

GURS: Orthophoto DOF025 (2017-2019), Free Access Database [online] Available from: www.gu.gov.si/en/services/free_access_database/ (Accessed 15 November 2020), 2020.

Hafner, P., McCarroll, D., Robertson, I., Loader, N. J., Gagen, M., Young, G. H. F., Bale, R. J., Sonninen, E. and Levanič, T.: A 520 year record of summer sunshine for the eastern European Alps based on stable carbon isotopes in larch tree rings,
455   Clim. Dyn., 43(3–4), 971–980, doi:10.1007/s00382-013-1864-z, 2014.

Haylock, M. R., Hofstra, N., Klein Tank, A. M. G., Klok, E. J., Jones, P. D. and New, M.: A European daily high-resolution gridded data set of surface temperature and precipitation for 1950-2006, J. Geophys. Res. Atmos., 113(20), doi:10.1029/2008JD010201, 2008.

Hernández, A., Kutiel, H., Trigo, R. M., Valente, M. A., Sigró, J., Cropper, T. and Santo, F. E.: New Azores archipelago daily
460   precipitation dataset and its links with large-scale modes of climate variability, Int. J. Climatol., 36(14), 4439–4454, doi:10.1002/joc.4642, 2016.

Herrera, S., Gutiérrez, J. M., Ancell, R., Pons, M. R., Frías, M. D. and Fernández, J.: Development and analysis of a 50-year high-resolution daily gridded precipitation dataset over Spain (Spain02), Int. J. Climatol., 32(1), 74–85, doi:10.1002/joc.2256, 2012.

465   Jevšenak, J.: Daily climate data reveal stronger climate-growth relationships for an extended European tree-ring network, Quat. Sci. Rev., 221, doi:10.1016/j.quascirev.2019.105868, 2019.

Jevšenak, J. and Levanič, T.: Using the dendroTools R package to analyse the relationships between daily climate and tree-ring data, , doi:10.13140/RG.2.2.30426.64966, 2018.

Jevšenak, J., Tychkov, I., Gričar, J., Levanič, T., Tumajer, J., Prislan, P., Arnič, D., Popkova, M. and Shishov, V. V: Growth-
470   limiting factors and climate response variability in Norway spruce (Picea abies L.) along an elevation and precipitation gradients in Slovenia, Int. J. Biometeorol., doi:10.1007/s00484-020-02033-5, 2020.

Kermavnar, J., Marinšek, A., Eler, K. and Kutnar, L.: Evaluating short-term impacts of forest management and microsite conditions on understory vegetation in temperate fir-beech forests: Floristic, ecological, and trait-based perspective, Forests, 10(10), doi:10.3390/f10100909, 2019.

475   Klanderud, K. and Totland, Ø.: Simulated climate change altered dominance hierarchies and diversity of an alpine biodiversity hotspot, Ecology, 86(8), 2047–2054, doi:10.1890/04-1563, 2005.

Klopčič, M., Poljanec, A., Dolinar, M., Kastelec, D. and Bončina, A.: Ice-storm damage to trees in mixed Central European forests: damage patterns, predictors and susceptibility of tree species, For. An Int. J. For. Res., 93(3), 430–443, doi:10.1093/forestry/cpz068, 2020.

480   Kozjek, K., Dolinar, M. and Skok, G.: Objective climate classification of Slovenia, Int. J. Climatol., 37, 848–860, doi:10.1002/joc.5042, 2017.

de Luis, M., Čufar, K., Saz, M. A., Longares, L. A., Ceglar, A. and Kajfež-Bogataj, L.: Trends in seasonal precipitation and

temperature in Slovenia during 1951–2007, Reg. Environ. Chang., 14(5), 1801–1810, doi:10.1007/s10113-012-0365-7, 2014.

485 Lussana, C., Tveito, O. E. and Uboldi, F.: Three-dimensional spatial interpolation of 2 m temperature over Norway, Q. J. R. Meteorol. Soc., 144(711), 344–364, doi:10.1002/qj.3208, 2018.

Martinez del Castillo, E., Prislan, P., Gričar, J., Gryc, V., Merela, M., Giagli, K., de Luis, M., Vavrčík, H. and Čufar, K.: Challenges for growth of beech and co-occurring conifers in a changing climate context, Dendrochronologia, 52, 1–10, doi:10.1016/j.dendro.2018.09.001, 2018.

490 Martínez del Castillo, E., Longares, L. A., Serrano-Notivoli, R. and de Luis, M.: Modeling tree-growth: Assessing climate suitability of temperate forests growing in Moncayo Natural Park (Spain), For. Ecol. Manage., 435, 128–137, doi:10.1016/j.foreco.2018.12.051, 2019.

Milošević, D. D., Savić, S. M., Stankov, U., Žiberna, I., Pantelić, M. M., Dolinaj, D. and Leščešen, I.: Maximum temperatures over slovenia and their relationship with atmospheric circulation patterns, Geogr. CGS, 122(1), 1–20,
495 doi:10.37040/geografie2017122010001, 2017.

Pogačar, T., Žnidaršič, Z., Črepinšek, Z. and Bogataj, L. K.: Aggravated Occupational Heat Stress Recognition and Mitigation in Slovenia, in Climate Change Adaptation in Eastern Europe, edited by W. Leal Filho, G. Trbic, and D. Filipovic, pp. 267–277, Springer, Cham., 2019.

Pogačar, T., Žnidaršič, Z., Bogataj, L. K. and Črepinšek, Z.: Steps towards comprehensive heat communication in the frame
500 of a heat health warning system in Slovenia, Int. J. Environ. Res. Public Health, 17(16), 1–16, doi:10.3390/ijerph17165829, 2020.

Prislan, P., Gričar, J., de Luis, M., Smith, K. T. and Čufar, K.: Phenological variation in xylem and phloem formation in Fagus sylvatica from two contrasting sites, Agric. For. Meteorol., 180, 142–151, doi:10.1016/j.agrformet.2013.06.001, 2013.

Serrano-Notivoli, R., Beguería, S., Saz, M. Á., Longares, L. A. and de Luis, M.: SPREAD: A high-resolution daily gridded
505 precipitation dataset for Spain, Earth Syst. Sci. Data Discuss., 1–33, doi:10.5194/essd-2017-35, 2017.

Serrano-Notivoli, R., Beguería, S. and de Luis, M.: STEAD: A high-resolution daily gridded temperature dataset for Spain, Earth Syst. Sci. Data Discuss., 1–27, doi:10.5194/essd-2019-52, 2019.

Škrk, N., de Luis, M., Čufar, K., Črepinšek, Z., Kajfež Bogataj, L. and Serrano-Notivoli, R.: SLOCLIM: A high-resolution daily gridded precipitation and temperature dataset for Slovenia [Dataset], , doi:10.5281/zenodo.4108543, 2020.

510 Vertačnik, G., Dolinar, M., Bertalanič, R., Klančar, M., Dvoršek, D. and Nadbath, M.: Ensemble homogenization of Slovenian monthly air temperature series, Int. J. Climatol., 35(13), 4015–4026, doi:10.1002/joc.4265, 2015.

Yatagai, A., Kamiguchi, K., Arakawa, O., Hamada, A., Yasutomi, N. and Kitoh, A.: Aphrodite constructing a long-term daily gridded precipitation dataset for Asia based on a dense network of rain gauges, Bull. Am. Meteorol. Soc., 93(9), 1401–1415, doi:10.1175/BAMS-D-11-00122.1, 2012.

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