ESTIMATION OF DAILY MEAN TEMPERATURES: AN ACCURATE METHOD FOR THE DOURO VALLEY

ESTIMAÇÃO DAS TEMPERATURAS MÉDIAS DIÁRIAS: UMA METODOLOGIA PRECISA PARA A REGIÃO DO DOURO

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SUMMARY

Air temperature data from many locations worldwide are only available as series of daily minima and maxima temperatures. Historically, several different approaches have been used to estimate the actual daily mean temperature, as only in the last two or three decades automatic thermometers are able to compute its actual value. The most common approach is to estimate it by averaging the daily minima and maxima. When only daily minima and maxima are available, an alternative approach, proposed by Dall’Amico and Hornsteiner in 2006, uses the two daily extremes together with next day minima temperature and a coefficient related to the local daily astronomical sunset time. Additionally, the method uses two optimize coefficients related to the region’s temperature profile. In order to use this approach it is necessary to optimize the region’s unknown parameters. For this optimization, it is necessary a dataset containing the maxima, minima, and the actual daily mean temperatures for at least one year. In this research, for the period 2007-2014, we used three datasets of minima, maxima and actual mean temperatures obtained at three automatic meteorological stations located in the Douro Valley to optimize the two unknown parameters in the Dall’Amico and Hornsteiner approach. Moreover, we compared the actual mean daily temperatures available from the three datasets with the correspondent values estimated by using i) the usual approach of averaging the daily maxima and minima temperatures and ii) the Dall’Amico and Hornsteiner approach. Results show that the former approach overestimates, on average, the daily mean temperatures by 0.5°C. The Dall’Amico and Hornsteiner approach showed to be a better approximation of mean temperatures for the three meteorological stations used in this research, being unbiased relative to the actual mean values of daily temperatures. In conclusion, this research confirms that the Dall’Amico and Hornsteiner is a better approach to estimate the mean daily temperatures and provides the optimized parameters for three sites located at each of the three sub-regions of the Douro Valley (Baixo Corgo, Cima Corgo and Douro Superior).

RESUMO

Os dados relativos à temperatura do ar, em muitos locais do mundo, estão apenas disponíveis na forma de séries diárias de temperaturas mínimas e máximas. Historicamente, várias abordagens diferentes têm sido usadas para estimar a temperatura média diária, dado que só nas últimas duas ou três décadas termômetros automáticos são capazes de calcular seu valor real. A abordagem mais comum, é estimá-la utilizando a média aritmética das temperaturas mínimas e máximas diárias. Quando apenas os mínimos e máximos diários estão disponíveis, uma abordagem alternativa, proposta em 2006 por Dall’Amico e Hornsteiner, usa os dois extremos diários juntamente com a temperatura mínima do dia seguinte e um coeficiente relacionado com a hora astronómica do pôr do sol, no local. Além disso, o método usa dois coeficientes otimizáveis, relacionados com o perfil de temperatura da região. Para usar essa abordagem é necessário otimizar os parâmetros desconhecidos da região. Para esta otimização, é necessário um conjunto de dados contendo os mínimos, mínimos e as temperaturas médias diárias reais de, pelo menos, um ano. Neste trabalho, para otimizar os dois parâmetros utilizados na abordagem proposta por Dall’Amico e Hornsteiner, no período de 2007 a 2014, foram utilizados três conjuntos de dados de mínimos, máximos e temperaturas médias reais, obtidas em três estações meteorológicas automáticas localizadas na Revião do Douro. Além disso, comparamos as temperaturas médias diárias reais, disponíveis em cada um dos três conjuntos de dados obtidos nas três estações meteorológicas, com os valores correspondentes, estimados usando i) a abordagem usual de média aritmética das temperaturas máximas e mínimas diárias e ii) a abordagem de Dall’Amico e Hornsteiner. Os resultados mostram que a primeira abordagem superestima, em média, as temperaturas médias diárias em 0.5°C. A abordagem de por Dall’Amico e Hornsteiner mostrou ser uma melhor aproximação das temperaturas médias nas três estações meteorológicas utilizadas neste trabalho, sendo não-enviesadas em relação aos valores médios reais das temperaturas diárias. Em conclusão, este trabalho confirma que a abordagem de Dall’Amico e Hornsteiner é melhor para estimar temperaturas médias diárias. São fornecidos os parâmetros otimizados para os três locais, situados em cada uma das três sub-regiões do Douro Vale (Baixo Corgo, Cima Corgo e Douro Superior).

Key words: daily mean temperature, monthly mean temperature, minimal temperature, maximal temperature, Douro Valley.

Palavras-chave: temperatura média diária, temperatura média mensal, temperatura mínima, temperatura máxima, região do Douro.

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INTRODUCTION

Temperatures gridded datasets start in 1850 because there are too few observations available from before this date (Brohan et al., 2006). However, local meteorological datasets exist since the 16th century as, for example, the 1756-1998 daily temperature and pressure dataset for Stockholm (Moberg et al., 2002). Until a few decades ago, temperatures were recorded from several manual daily readings. For the Stockholm dataset, during 1756–1760, the temperature readings were made around sunrise and 13:00; during 1761-1783 the temperature readings were made around 6:30-8:00, 13:00 and 22:00-23:00; and in 1784 observations at 06:00, 14:00 and 21:00 were introduced for all variables. At the present time, effective practices for carrying out meteorological observations and measurements of temperature according to internationally agreed standards may be found in (World Meteorological Organization, 2014). Presently, there is still not a widespread consensus among meteorological organizations relative to the definition of mean daily temperature. For example the World Meteorological Organization defines the mean daily temperature as “Mean of the temperatures observed at 24 equidistant times in the course of a continuous interval of 24 hours (normally the mean solar day, from midnight to midnight, according to the zonal time or the mean solar time of the station); or a combination of temperatures observed at less numerous times, so arranged as to depart as little as possible from the mean defined above” (World Meteorological Organization, 2016). On the other hand, the Government of Canada defines the mean daily temperature as “The average of the maximum and minimum temperature at a location for a specified time interval” (Government of Canada, 2016).

Daily mean temperatures are used as input variables in a number of climatological, physical, ecological, agricultural, biological, and technical processes and applications. Historically, several different approaches have been used to estimate the actual daily mean temperature as it was only in the last two or three decades that automatic thermometers that operate continuously are able to compute its mathematical value: \[
T = \frac{1}{24} \int_{00:00}^{23:59:59} T \, dT
\]

Several approaches estimate the daily mean temperature by averaging a collection of several daily temperature readings taken at different times of the day (Sakellariou and Kambezidis, 2016; Ma and Guttorp, 2013; Weiss and Hays, 2005).

In many countries and in many places, the daily mean temperature still is estimated using the common approach of averaging the maxima and minima temperatures of a day. In this research, this approach will be called the TAvg method. It is long known that the use of this method produces poor estimates of the actual mean daily temperature Kaemtz et al. (1845).

In the Douro Valley (see Figure 1), meteorological and climatological studies about its present and future climate, as well as agricultural studies on the region’s crops and agricultural products (grapes, olives, almonds, figs, oranges and cherries), use estimations of mean daily temperatures. Most of these studies have used the TAvg method to estimate the unknown real mean temperature (Santos and Malheiro, 2011; Jones, 2012; Jones and Alves, 2012; Jones 2013; Fraga et al., 2014; C. Real et al., 2014; C. Real et al., 2016; Fraga and Santos, 2017).

![Figure 1. Portugal and the location of the Douro Valley (relief maps: www.maps-for-free.com).](image1)

An alternative estimation approach, proposed by Dall’Amico and Hornsteiner (2006) that we will call the Dall’Amico method, uses the two daily extreme temperatures together with the next day minima temperature and a coefficient related to the local daily astronomical sunset time (US Naval Observatory, 2017). Additionally, the method uses two optimizable coefficients, related to the temperatures profiles of the region. For the optimization of the two unknown parameters the authors suggest that a dataset containing the maxima, minima, and the actual daily
mean temperatures for at least one year should be used to fit the coefficients to the available data.

This research used three datasets of maxima, minima and mean daily temperatures collected at three meteorological stations located at each of the three sub-regions of the Douro Valley (Baixo Corgo, Cima Corgo and Douro Superior – see Figure 2). As these datasets were obtained in automatic meteorological stations that are able to measure temperatures continuously, the mean temperatures in the datasets are the actual values. We compared the actual mean daily temperatures to the estimates obtained by using the TAvg method and by using the alternative method proposed by Dall’Amico and Hornsteiner (2006). The quality of the estimates obtained by the two methods was assessed.

Figure 2. Location of the local meteorological stations.

We are aware that the use of simple methods as the one proposed by Dall’Amico and Hornsteiner are prone to lead to poorer results when used in sites where a significant spatial variability exists. For these sites, local landscape may determine highly variable patterns for the evolution of temperatures along each day. This is the case of mountain regions such as the Douro Valley, interrupted longitudinally by the Douro river valley, where mesoscale weather processes, the existence of dynamically and thermally wind systems, sunshine exposition, clouds at site level and other factors, influence the daily evolution of the weather at the site. Nevertheless, after having the optimized coefficients for the Dall’Amico and Hornsteiner expression its use is intended to estimate the unknown real value of the mean temperature, using only the daily extreme temperatures.

**MATERIAL AND METHODS**

**Data**

Three datasets of maxima, minima and mean daily temperatures for the period 2007-2014, collected at Cambres (91 m elevation), Pinhão (107 m elevation) and Vilariça (171 m elevation) stations, located at each of the three sub-regions of the Douro Valley (Baixo Corgo, Cima Corgo and Douro Superior), have been used (Figure 2). Considering a constant lapse rate of 5.1 °C/km(elevation) (Stone and Carlson 1979), the difference in temperatures from station to station that may be inputable to the difference in the elevation is less than 0.4 °C. A methodology developed by Feng et al. (2004) was used to identify data that was suspect of being erroneous. In order to homogenize the cleaned dataset the RHtestsV3 software package (Wang, 2011) was used.

Table I shows the maximum, minimum and mean temperatures at Cambres, Pinhão and Vilariça for each quarter in 2007-2014. Cambres, Pinhão and Vilariça stations are located at each of the three sub-regions of the Douro Valley (Baixo Corgo, Cima Corgo and Douro Superior). Vilariça is the station that is located further inland and Cambres is the station that is closer to the Atlantic Ocean coast. Analysis of the minimum temperatures in each station at the four quarters show that during the coldest months, minimum temperatures decrease from the coast towards inland and maximum temperatures are similar in the three stations. During the warmest months, minimum temperatures are similar in the three stations and maximum temperatures increase from the coast towards inland.

**Methods**

Estimates of mean temperatures, obtained by using two methods of estimation (TAvg and Dall’Amico), were compared with the correspondent real mean temperatures.

**T Avg method**

The TAvg method is the most common and simple method for estimating the mean daily temperature. The mean daily temperature is estimated by averaging minima and maxima daily temperatures, see equation 1.

\[
\bar{T}_{\text{mean}}^i = \frac{T_{\text{max}}^i - T_{\text{min}}^i}{2}
\]  

(1)
Table I
Temperatures at Cambres, Pinhão and Vilariça (2007-2014)

| Station | months | Tmean | Tmax | Tmin | Tmean | Tmax | Tmin | Tmean | Tmax | Tmin | Tmean | Tmax | Tmin |
|---------|--------|-------|------|------|-------|------|------|-------|------|------|-------|------|------|
| Cambres elevation: 91m |
| 1-3     | 10.2   | 15.3  | 5.9  | 10.7 | 16.1  | 6.5  | 10.4 | 16.7  | 5.2  | 9.4  | 14.0  | 5.8  |
| 4-6     | 17.6   | 24.8  | 11.3 | 17.5 | 24.1  | 11.6 | 17.9 | 25.0  | 11.6 | 18.3 | 25.7  | 11.8 |
| 7-9     | 22.1   | 30.7  | 14.3 | 21.9 | 29.5  | 14.9 | 23.3 | 31.7  | 15.8 | 24.3 | 33.1  | 16.7 |
| 10-12   | 10.3   | 18.2  | 5.0  | 10.6 | 16.7  | 5.9  | 13.1 | 17.9  | 9.5  | 11.0 | 16.4  | 7.0  |
| Pinhão elevation: 107m |
| 1-3     | 10.0   | 14.6  | 5.9  | 10.2 | 15.4  | 6.3  | 10.0 | 16.2  | 5.1  | 9.2  | 13.9  | 5.7  |
| 4-6     | 17.8   | 24.2  | 12.0 | 17.5 | 24.1  | 12.2 | 18.4 | 25.2  | 12.1 | 18.3 | 25.1  | 12.1 |
| 7-9     | 23.0   | 30.7  | 15.9 | 22.5 | 29.8  | 15.7 | 24.0 | 31.9  | 16.7 | 25.0 | 33.4  | 17.6 |
| 10-12   | 10.3   | 16.9  | 5.8  | 10.5 | 15.9  | 6.4  | 13.2 | 17.9  | 9.6  | 11.0 | 16.1  | 7.3  |
| Vilariça elevation: 171m |
| 1-3     | 9.5    | 14.8  | 4.3  | 9.6  | 15.5  | 4.8  | 9.6  | 16.4  | 3.7  | 8.8  | 13.7  | 4.6  |
| 4-6     | 17.7   | 24.8  | 10.8 | 17.8 | 24.8  | 11.2 | 19.0 | 26.0  | 11.7 | 18.0 | 25.5  | 10.9 |
| 7-9     | 23.5   | 32.0  | 15.0 | 23.3 | 31.7  | 15.1 | 25.0 | 33.4  | 16.7 | 25.6 | 34.7  | 17.6 |
| 10-12   | 9.3    | 17.1  | 3.3  | 9.6  | 15.7  | 4.6  | 12.9 | 17.9  | 8.6  | 10.3 | 16.1  | 5.5  |

Dall'Amico method

Dall’Amico method (Dall’Amico and Hornsteiner 2006) is a method for estimating the mean daily temperature, Tmean, on the basis of minima and maxima daily temperatures. The method accounts for the temperature trend by including the minimum temperature on the following day in addition to the extremes on the day in question, see equation 2.

\[
T_{\text{mean}} = \left[ T_{\text{min}} + C_D (T_{\text{max}} - T_{\text{min}}) \right] S_{\text{day}}^i + \left[ T_{\text{min}} + C_N (T_{\text{max}} - T_{\text{min}}) \right] (1 - S_{\text{day}}^i) \tag{2}
\]

Where:

* \( S_{\text{day}}^i \) is the nondimensional astronomical sunset time on day i, \( S_{\text{day}}^i = \frac{t_{\text{day}}^i}{24} \) where \( t_{\text{day}}^i \) is the astronomical sunset time on day i.

The astronomical sunset times for any location are available on the internet: e.g. [http://aa.usno.navy.mil/data/docs/RS_OneYear.php](http://aa.usno.navy.mil/data/docs/RS_OneYear.php);

* \( C_D \) is the proportionality coefficient for the day;

* \( C_N \) is the proportionality coefficient for the night.

As suggested in Dall’Amico and Hornsteiner (2006) seasonally varying coefficients \( C_D \) and \( C_N \) in equation 2 have been optimized partitioning years into two periods: one that will be designated as cold period and the other as warm period, \( C_D^{\text{Warm}} \), \( C_D^{\text{Cold}} \), \( C_N^{\text{Warm}} \), and \( C_N^{\text{Cold}} \). The optimization consisted on the minimization of the Root Mean Square Errors (RMSE) of the estimates for the mean daily temperatures at Cambres, Pinhão and Vilariça stations obtained by using equation 2, relative to their actual mean daily temperatures. Parameter optimization for the Dall’Amico method was performed by using the Microsoft Excel (2016) evolutionary solver algorithm. The optimization process has also determined the boundaries of the two periods (cold and warm).

\[1\] Dall’Amico and Hornsteiner (2006) have tested their methodology in two different sites: one located in a European flat terrain and the other located in an Alpine valley site. They concluded that the use of the actual sunset times yielded slightly worse results when compared to the astronomical sunset time.
For both methods, the respective Root Mean Square Errors (RMSE) and bias have been calculated in order to assess their relative accuracy.

Assessing model for overfitting

To assess if the estimates obtained by using the Dall’Amico method were overfitting the data, we evaluated the model by using a cross-validation methodology. To that effect, 4-fold methodology was defined. The datasets of maxima, minima and mean daily temperatures from eight years (2007-2014) were separated into four groups (folds) having each group the data from two consecutive years. Six years of data (tree folds) were used to estimate the model parameters (training the model) and the two remaining years of data (one fold) were used to validate the accuracy of the model (testing the model).

Additionally, in order to assess if the model parameters would maintain valid for a period of seven years we evaluated the quality of the estimates obtained using a single year of data to compute the model parameters (training set) and the remaining seven years of data to test the model’s estimates (testing set). We have done this evaluation using each single year of data to compute the model parameters (training set). We will present results naming this methodology as “validation 1/7”.

RESULTS AND DISCUSSION

The values for the boundaries of the cold and warm periods resulted from the optimization process: warm period: March 1 to September 30 and cold period: January 1 to February 28 and from October 1 to December 31. Additionally, four optimum values for the seasonally varying proportionality coefficients were obtained for the Douro Valley, which are presented in Table II.

Using each of the two pairs of $C_i$ ($C_D^{\text{Warm}}, C_N^{\text{Warm}}$ or $C_D^{\text{Cold}}, C_N^{\text{Cold}}$) in equation 2, two forms of this equation were obtained. The first, uses $C_D^{\text{Warm}}, C_N^{\text{Warm}}$, and is intended for the estimation of the mean daily temperature for the warm months (months 3 to 9). The second, uses $C_D^{\text{Cold}}, C_N^{\text{Cold}}$, and is intended for the estimation of the mean daily temperature for the cold months (months 1, 2, 10, 11, and 12).

The Dall’Amico Method estimator is unbiased while the TAvg method is a biased estimator of the actual mean daily temperature, overestimating the actual value, on average, 0.5°C.

We assessed the accuracy of the mean temperatures estimates calculated using the Dall’Amico Method when using coefficients $C_D$ and $C_N$ optimized for a site other than the one for which the estimates are computed. We run several combinations using coefficients $C_D$ and $C_N$ optimized using the data from one single site and then assessing, for the two remaining sites, the accuracy of the estimates computed using the previously optimized $C_D$ and $C_N$. The corresponding bias and RMSE have been computed and the results are shown in Table IV.

The results of the three combinations in Table IV show that when compared to the TAvg method, the Dall’Amico Method has comparable values for RMSE and a smaller bias. In a small region such as
the Douro Valley, despite the fact that landscape may determine highly variable patterns for the evolution of temperatures along each day, the results indicate that the use of the optimized coefficients presented in the "the 3 stations" column in Table II will lead to estimates of the mean daily temperatures having a comparable variability but a smaller bias, relative to the TAvg method.

| Table IV |
|-------------------------|
| Bias and RMSE computed using coefficients $C_D$ and $C_N$ optimized for a single site |

**Enviesamento e RMSE para as estimativas calculadas com os coeficientes $C_D$ e $C_N$ optimizados para uma só estação**

Using the Dall’Amico Method with the $C_D$ and $C_N$ coefficients optimized using temperatures data from the three stations, we compared the differences between the actual mean monthly temperatures (the average of all mean daily temperatures during the correspondent month) and its estimates obtained by the two methods (TAvg and Dall’Amico) at each station (see Figure 3, Figure 4 and Figure 5).

In the 96-month period (Jan 2007 to Dec 2014), in the three stations (Cambres, Pinhão and Vilarica), 79.51% of the mean monthly temperatures estimated using the Dall’Amico method have a smaller deviation to the actual monthly mean temperatures than the estimates obtained using the TAvg method.

These results, together with the results obtained using mean daily temperatures (see Table III), show that the Dall’Amico method is a better approximation for both daily and monthly mean temperatures, having smaller bias and RMSE relative to the estimates obtained using the TAvg method.

**Figure 3.** Differences from the estimated mean monthly temperatures to its actual value at Cambres. A constant offset of 0.5°C exists, on average, between the values of the two estimation methods. Bias Dall’Amico Method equals 0.1. Bias TAvg method equals 0.6.

**Diferença entre os valores estimados para as temperaturas médias mensais em Cambres e o seu valor real. Em média, verifica-se um desfasamento de 0.5°C entre os dois métodos de estimação. O enviesamento das temperaturas estimadas pelo método de Dall’Amico & Hornsteiner é de 0.1° e o enviesamento das temperaturas estimadas pelo método TAvg é de 0.6°.**
Results of the cross-validation methodology used to assess model overfitting are presented in Table V and they show that the complete model has a RMSE and a bias comparable to the correspondent values obtained by averaging the 4-fold individual results indicating that the complete model is stable when used with new unknown data.

Results for the methodology validation 1/7 are presented in Table VI showing that even when using a single year of data to train the model, the obtained coefficients enable equation 2 to produce good estimates of the mean daily temperatures when using new unknown data as input.

These results suggest that for a slow climate change process, the obtained coefficients will only require sparse recalibrations maintaining useable during periods of 7 or more years. Although the average of the eight resulting partial models has an RMSE and a bias both larger than correspondent values of the complete model, the estimates are still better than the ones obtained by using the TAvg method.
The present research shows that the TAvg estimation method overestimates the actual mean daily air temperatures in the Douro Valley. On average, the overestimation is of 0.5ºC, being larger for the warm months (0.6ºC) than for the remaining months (0.3ºC). The alternative estimation method, the Dall’Amico method, is also a simple method for estimating the mean daily temperature based on the daily maxima and minima temperatures. This latter method showed to be closer to the actual values of the mean daily temperature for every month of the year, having a small bias relative to the actual values (0.0 for daily mean temperatures and 0.13 for monthly mean temperatures in the 96-month period analyzed in this research).

Table V
Results for the cross-validation methodology

| Fold | Dall’Amico RMSE | TAvg RMSE | Fold 1 | Fold 2 | Fold 3 | Fold 4 | Average 4 folds | Complete model |
|------|-----------------|-----------|-------|-------|-------|-------|-----------------|----------------|
| 1    | 0.6600          | 0.8600    | 0.6500| 0.6500| 0.7000| 0.6000| 0.6525          | 0.6500         |
| 2    | -0.0444         | 0.0234    | 0.0148| 0.0122| 0.0015| 0.0021|                 |                |
| 3    | 0.4321          | 0.4870    | 0.4750| 0.4536| 0.4619| 0.4619|                 |                |

Table VI
Results for the approach of using a single year of data to estimate the model parameters

| Training year | 2007 | 2008 | 2009 | 2010 | 2011 | 2012 | 2013 | 2014 | Average | Complete model |
|---------------|------|------|------|------|------|------|------|------|---------|----------------|
| Dall’Amico RMSE | 0.6600 | 0.6600 | 0.6600 | 0.6700 | 0.6700 | 0.6600 | 0.6600 | 0.6600 | 0.6633 | 0.6500         |
| bias           | 0.0030 | 0.0460 | 0.0687 | -0.0412 | -0.0744 | 0.0501 | -0.0081 | -0.1168 | -0.0203 | 0.0021         |
| TAvg RMSE      | 0.8500 | 0.8600 | 0.8600 | 0.8700 | 0.8500 | 0.8600 | 0.8600 | 0.8400 | 0.8567 | 0.8600         |
| bias           | 0.4600 | 0.4663 | 0.4724 | 0.4625 | 0.4504 | 0.4647 | 0.4637 | 0.4555 | 0.4615 | 0.4619         |

Several meteorological and agricultural indexes based on daily temperatures use mean daily temperatures as input. It is not always clear for which methodology of mean temperatures calculation they are calibrated to. In viticulture, for example, growing degree-days (GDD) are defined as the sum of the mean daily temperature above a threshold from January 1 to a given date:

\[
GDD = \sum_{i=1}^{\text{Date}} (T_{\text{avg}} - T_{\text{base}})
\]

\(T_{\text{avg}}\) is the average temperature and \(T_{\text{base}}\) is a temperature used as threshold. The \(T_{\text{avg}}\) value used in the calculation of GDD is not the actual mean daily temperature but its estimation obtained by using the TAvg method. For example, van Leeuwen et al. (2008) studied the heat requirement (GDD) for each grapevine variety to reach each phenological event defining heat requirement data for most grape varieties. Using daily average temperatures obtained by a methodology that produces estimates closer to the unknown daily mean temperatures than the TAvg estimates will lead to a definition of more precise heat requirements for the grapevine phenological events. Nowadays, however, more accurate and consistent growing heat summation indices such as growing degree hours (GDH) based on hourly temperatures, are replacing GDD for phenological prediction (Gu, 2016).

Optimized parameters for three sites located at each of the three sub-regions of the Douro Valley (Baixo Corgo, Cima Corgo and Douro Superior), for the period 2007-2014, are made available in this research. Interpolation of the values of the parameters will allow their use to estimate mean temperatures (with moderate confidence) in sites different from the ones used in the optimization process. As the three stations are located at sites having an average elevation of 120 meters, the use of the optimized parameters made...
available in this research will make necessary to refer the daily maxima and minima to a reference elevation of 120. Rolland (2003) indicates that temperature decreases, on average, 5 °C / 1000 meters elevation increase.

The use of more precise methods for the estimation of the actual mean daily temperature, although providing estimates closer to the actual values, should be followed by a recalibration of the indices and variables dependent on mean temperatures because, as in the GDD example, the indicative values of some indices are based on estimates obtained by using the routine TAvg method.

CONCLUSIONS

In the Douro Valley, for the period of 2007-2014 and for the three meteorological stations used in this research, the common approach overestimates, on average, the daily mean temperatures by 0.5 °C (0.90 overestimates, 0.06 underestimates, 0.04 exact matches) while the Dall’Amico method has, in the same period, a bias of 0.0 °C (0.44 overestimates, 0.40 underestimates, 0.16 exact matches). Although this does not seems a large bias, it will be interesting to note that in 1980-2009, the mean annual temperature in the Douro Valley increased 1.0 °C, from an average value of 14.9 °C in the early 80s to 15.9 °C at 2009 (Sousa 2015). Estimates which overestimate the actual mean temperature, on average, by 0.5 °C, correspond, if the trend in 1980-2009 maintains, to the actual mean temperatures that will happen in this region by 2032, 15 years from now.

In conclusion, this research confirms that the Dall’Amico method should be preferred to estimate the mean daily temperatures.

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