

VALIDATION AND COMPARISON OF CLIMATE CHANGE SCENARIO STATISTICAL DOWNSCALING METHODS



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ABSTRACT

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Climate change impacts are very dependent on regional geographical features, local climate variability, and socio-economic conditions. Impact studies should therefore be performed at the local or at most at the regional level. However, climate scenarios are produced by Global Circulation Models (GCMs) for the entire globe with spatial resolutions of several hundred kilometres. Downscaling methods are thus needed to bridge the gap between the large scale climate scenario datasets and the fine scale where local health impacts happen.

In this study two statistical downscaling methods were assessed to see their ability in representing local climate variability. An inter-comparison and uncertainties analysis study between a stochastic weather generator (LARS-WG) and a hybrid of stochastic weather generator with transfer function methods (SDSM) was performed. Models errors and uncertainties were estimated using non-parametric statistical methods at the 95% confidence interval for precipitation, maximum temperature and minimum temperature for the mean and variance for a single site in Lisbon, Portugal.

The comparison between the observed dataset and the simulations showed that both models performance are acceptable. However, the SDSM tool was able to better represent the minimum and maximum temperature while LARS-WG simulations on precipitation are slightly better. The analysis of both models uncertainties for the mean are very close to the observed data in all months, but the uncertainties for the variances showed that the LARS-WG simulation performance is slightly better for precipitation and that both model simulations for minimum and maximum temperature are very close from the observed.

METHODS

Two different statistical downscaling methods were tested for a single site in Lisbon, Portugal. The first method is a stochastic weather generator that can be used for the simulation of weather for a single site under current and future climate conditions. This method was applied using the LARS-WG (Long Ashton Research Station Weather Generator) tool developed by Semenov and Brooks, (1999) that provides means of simulating synthetic weather time series with similar statistical characteristics of the observed statistics at a site.

The second statistical downscaling method is a hybrid of stochastic weather generator with transfer function methods. In this case, large-scale circulation patterns and atmospheric moisture variables were used to condition local scale weather parameters such as precipitation occurrence and intensity. Stochastic techniques are then used to artificially inflate the variance of the downscaled daily time series. This method was applied using the SDSM (Statistical DownScaling Model) tool developed by Wilby, Dowson and Barrow, 2001.

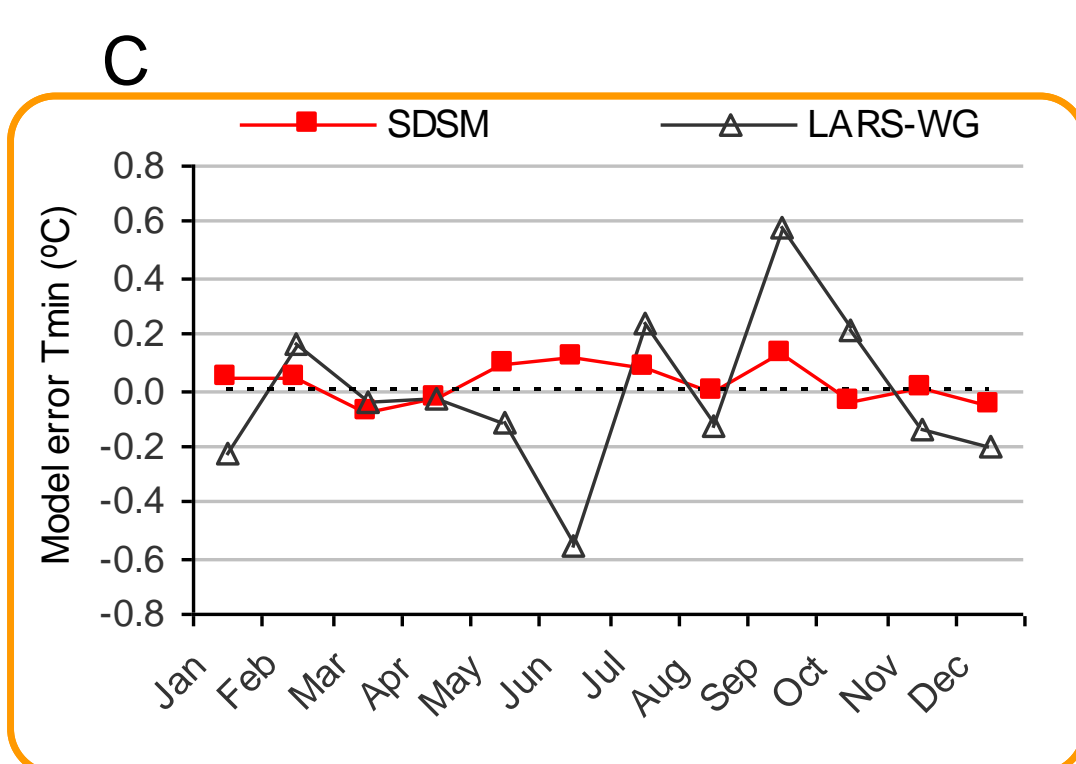
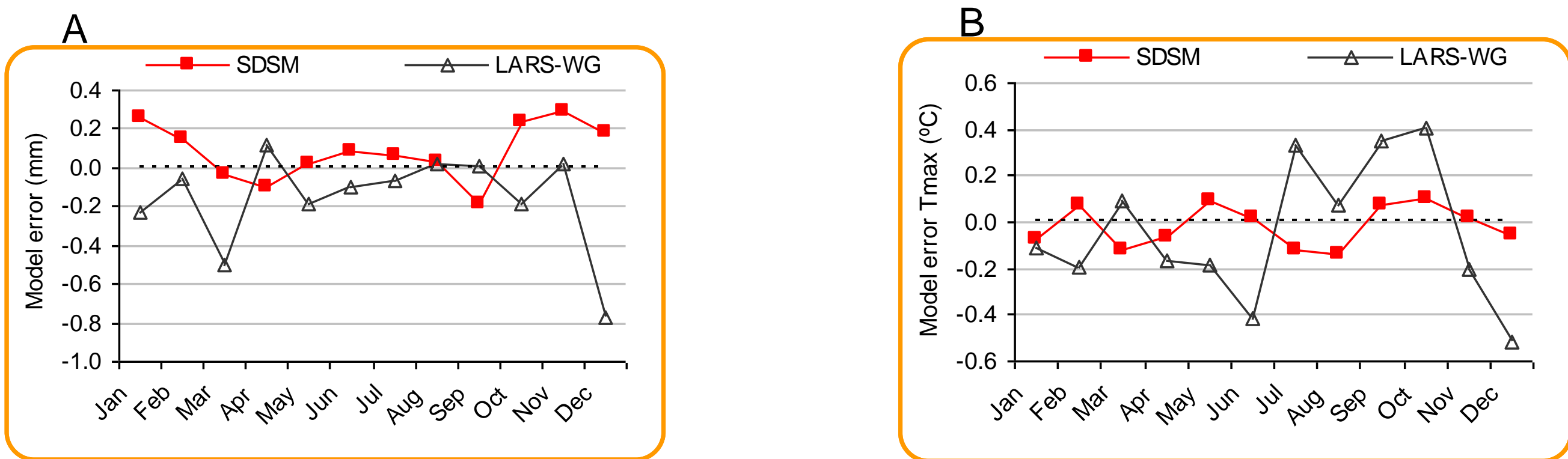
The observed climatic inputs for both statistical methods was daily precipitation, daily maximum temperature and daily minimum temperature for Lisbon during the 1961-1990 period, obtained from the European Climate Assessment & Dataset (ECA&D).

The GCM chosen was the coupled atmosphere-ocean HadCM3 developed by the Hadley Centre, with a horizontal resolution of 2.5° of latitude and 3.75° of longitude producing a global grid of 96 x 73 grid cells. For this study the GCM HadCM3 daily data were collected in two different sources. The observed predictors and the predictors for the A2 scenario used in the SDSM tool were acquired at the Canadian Climate Change Scenarios Network (CCSN), and the input for the LARS-WG tool were collected at the British Atmospheric Data Centre, from the Climate Impacts LINK Project.

Model validation of the downscaled daily precipitation and daily maximum and minimum temperature (PP, Tmax and Tmin respectively) was assessed in terms of model errors in the estimates of the mean and variances for the 1981-1990 period using non-parametric statistical analysis tests at 95% confidence level. For the mean we used the non-parametric Mann-Whitney U test that constructs the hypothesis test p value for the difference of the two population means ($\mu_1 - \mu_2$), assessing whether two samples of observations come from the same distribution (Mann and Whitney, 1947). The null hypothesis is that there is no difference between the two population means. For $p < 0.05$ the null hypothesis is rejected at 95% confidence level and the two populations means are not equal, the null hypothesis is accepted when $p > 0.05$. The chosen non-parametric test for the equality of two population variances was the Brown-Forsythe test (Brown and Forsythe, 1974). The 95% confidence intervals were calculated to assess the uncertainty in the estimate of means and variances using a bootstrapping non-parametric approach (Mooney and Duval, 1993). Bootstrapping is a method of estimating the properties of an estimator, such as the mean and variance, by measuring its properties when sampling from an approximating distribution.

RESULTS

Evaluation of the errors in the estimates of means:

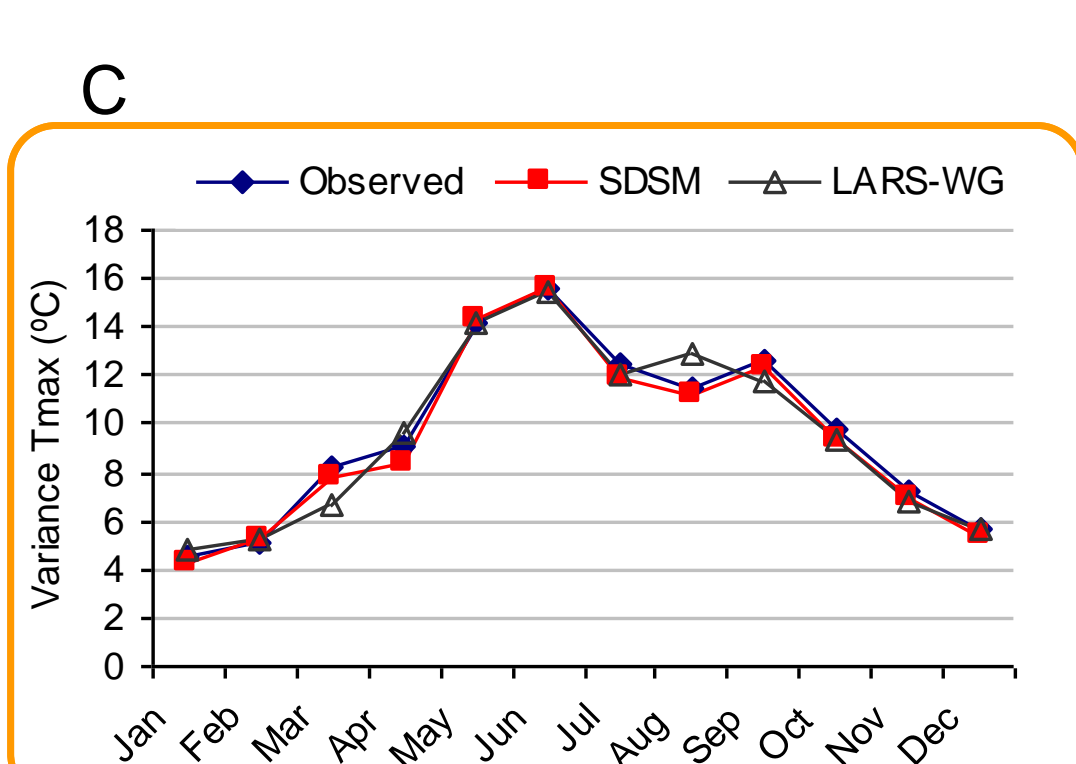
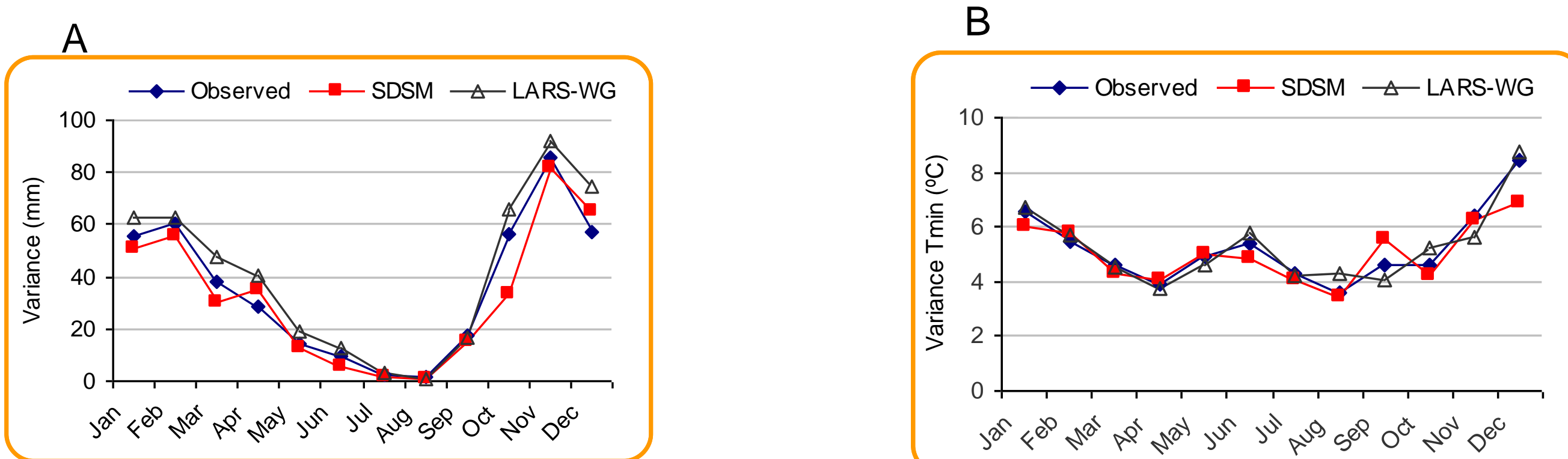


Observed minus simulated monthly means of (A) precipitation; (B) minimum temperature and (C) maximum temperature, for the 1981-1990 period.

	Tmin		Tmax		Precipitation	
	SDSM	LARS-WG	SDSM	LARS-WG	SDSM	LARS-WG
Jan	0.828	0.039	0.828	0.796	0.014	0.076
Feb	0.689	0.078	0.593	0.215	0.020	0.995
Mar	0.433	0.682	0.047	0.947	0.112	0.007
Apr	0.575	0.500	0.101	0.206	0.138	0.002
May	0.596	0.176	0.303	0.065	0.327	0.813
Jun	0.825	0.000	0.382	0.002	0.151	0.862
Jul	0.098	0.995	0.034	0.341	0.323	0.525
Aug	0.069	0.011	0.033	0.418	0.623	0.656
Sep	0.561	0.000	0.663	0.153	0.826	0.460
Oct	0.502	0.032	0.341	0.024	0.215	0.360
Nov	0.715	0.216	0.691	0.035	0.012	0.372
Dec	0.426	0.103	0.771	0.000	0.025	0.053

Test results (p values) of the Mann-Whitney U test for the difference of means of the observed (1981-1990) and downscaled daily Tmin, Tmax and precipitation at the 95% confidence level

Evaluation of the errors in the estimates of variances:

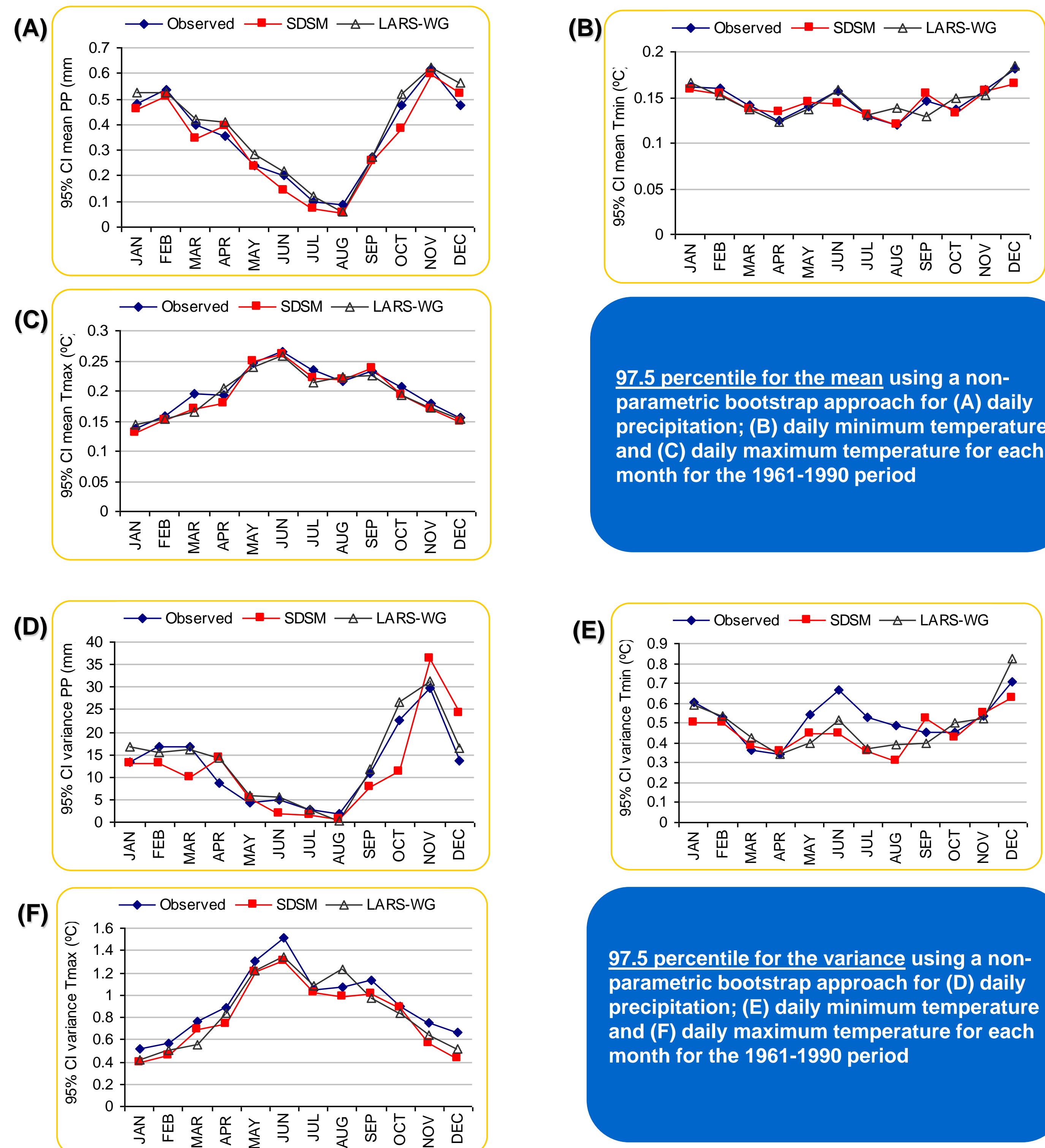


Estimation of the monthly average of daily observed and simulated variances for (A) precipitation; (B) maximum temperature and (C) minimum temperature for the 1981-1990 time period

	Tmin		Tmax		Precipitation	
	SDSM	LARS-WG	SDSM	LARS-WG	SDSM	LARS-WG
Jan	0.167	0.591	0.771	0.003	0.441	0.523
Feb	0.261	0.482	0.033	0.158	0.705	0.893
Mar	0.224	0.418	0.735	0.045	0.905	0.099
Apr	0.326	0.771	0.509	0.240	0.696	0.685
May	0.265	0.792	0.695	0.363	0.917	0.318
Jun	0.844	0.065	0.674	0.764	0.523	0.513
Jul	0.164	0.125	0.530	0.505	0.310	0.383
Aug	0.244	0.000	0.687	0.114	0.563	0.771
Sep	0.003	0.110	0.853	0.397	0.336	0.967
Oct	0.133	0.147	0.495	0.728	0.449	0.614
Nov	0.701	0.028	0.470	0.832	0.500	0.965
Dec	0.002	0.979	0.483	0.565	0.613	0.042

Test results (p values) of the Brown-Forsythe test for the difference of variances of the observed and downscaled daily Tmin, Tmax and precipitation at the 95% confidence level

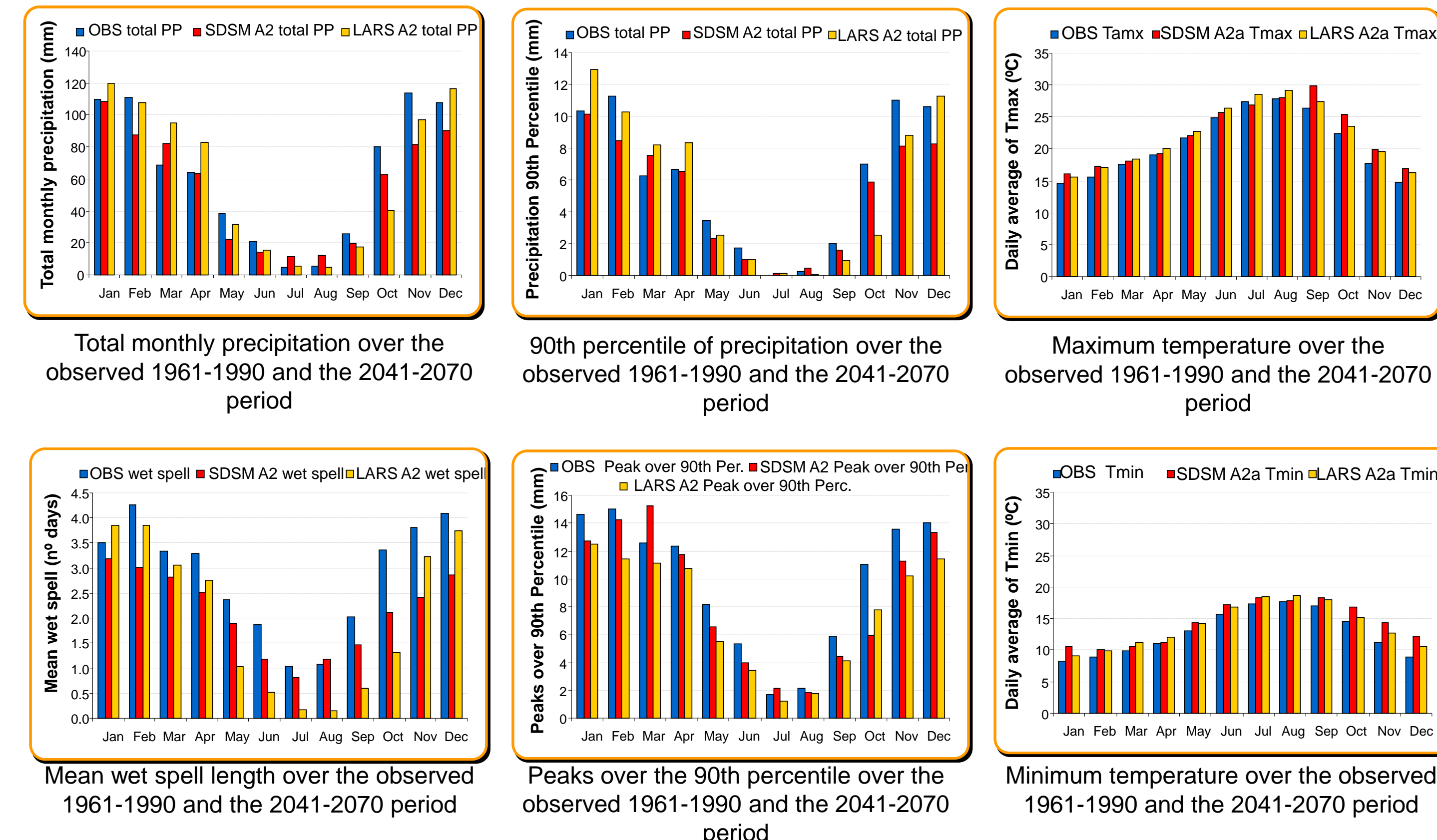
Uncertainties analysis for the mean and variance:



97.5 percentile for the mean using a non-parametric bootstrap approach for (A) daily precipitation; (B) daily minimum temperature and (C) daily maximum temperature for each month for the 1961-1990 period

97.5 percentile for the variance using a non-parametric bootstrap approach for (D) daily precipitation; (E) daily minimum temperature and (F) daily maximum temperature for each month for the 1961-1990 period

Summary statistics for Lisbon using LARS-WG and SDSM for the A2a SRES scenario:



DISCUSSION AND CONCLUSION

The SDSM and LARS-WG validation and model comparison was done using non-parametric statistics at the 95% confidence interval for daily precipitation, daily maximum temperature and daily minimum temperature for the mean and variance.

The model errors for the mean were tested using the Mann-Whitney U test and the performance of both model simulations are acceptable but the SDSM tool can better represent minimum and maximum temperature while LARS-WG simulations on precipitation are better. It is also important to note that the SDSM simulation for minimum temperature for the mean was well represented in all months. The Brown-Forsythe test was used to compare the difference of variances of the downscaled dataset. Both model performances were very good in almost all months.

In conclusion model errors and variability are close to the observed dataset for precipitation, maximum temperature and minimum temperature.

The general analysis of the simulations for the A2a SRES scenario for the 2041-2070 period showed that both techniques can be successfully used to produce downscaling scenarios at a local scale.

REFERENCES

Brown MB, Forsythe AB (1974) Robust Tests for Equality of Variances. Journal of the American Statistical Association 69:364-367
 Mann, H. B., & Whitney, D. R., 1947. On a test of whether one of two random variables is stochastically larger than the other. Annals of Mathematical Statistics, 18, 50-60.
 Wilby RL, Dawson CW, Barrow EM (2002) SDSM - a decision support tool for the assessment of regional climate change impacts. Environmental Modelling & Software 17:147-159
 Semenov MA, Brooks RJ (1999) Spatial interpolation of the LARS-WG stochastic weather generator in Great Britain. Climate Research 11:137-148