

Study of the impact of climate change on precipitation in Paris area using a method based on iterative multi scale dynamic time warping (IMs-DTW)

Djallel Dilmi, Laurent Barthès, Cécile Mallet, Aymeric Chazottes

▶ To cite this version:

Djallel Dilmi, Laurent Barthès, Cécile Mallet, Aymeric Chazottes. Study of the impact of climate change on precipitation in Paris area using a method based on iterative multi scale dynamic time warping (IMs-DTW). 9th International Workshop on Climate Informatics, Oct 2019, Paris, France. insu-02314175

HAL Id: insu-02314175 https://insu.hal.science/insu-02314175

Submitted on 11 Oct 2019 $\,$

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.



universite

PARIS-SACLAY

STUDY OF THE IMPACT OF CLIMATE CHANGE ON PRECIPITATION IN PARIS AREA USING METHOD BASED ON ITERATIVE MULTISCALE DYNAMIC TIME WARPING (IMS-DTW)

M. D. Dilmi*, L. Barthes, C. Mallet, A. Chazottes



LATMOS/IPSL, UVSQ Université Paris-Saclay, UPMC Univ. Paris 06, CNRS, Guyancourt, France *djallel.dilmi@latmos.ipsl.fr

0.ABSTRACT

Change on precipitation is constrained by finding a way to evaluate the evolution of precipitation variability over time. Classical approaches (feature-based) have shown their limitations for this issue due to the intermittent and irregular nature of precipitation. In this study, we present a novel variant of the Dynamic time warping method quantifying the dissimilarity between two rainfall time series based on shapes comparisons, for clustering annual time series recorded at daily scale. This shape-based approach considers the whole information (variability, trends and intermittency...). We further labeled each cluster using a feature-based approach. While testing the proposed approach on the time series of Paris-Montsouris, we found that the precipitation variability increased over the years in Paris area.

MOTIVATION



Iterative Multiscale Dynamic Time Warping(IMs-DTW)



dissimilarity scores	- This o by the	offsets w bservatio e IMS-DTV	ith up to on is true N betwee	two week for all th en the 14	ks. e alignme 6 times se	ents prop eries.	osed	time series 1890 to 1900		0.00 2018 2018 2018 2018 2018 2018 2019 0.00 0.00 0.00 0.00 0.00 0.00 Fig. the dissimilarity matrix D (146x146) between the 146 annual time series measured in Paris-Montsouris between 1873 and 2019		
Perform K-medoids clustering on annual time series based on dissimilarity scores	" k-me partition clusterin data se clusters, clusters priori a object to for each	edoids is ning te ng, which of n of with the assume and choos o the cen cluster "	s a cla echnique h cluster objects i e numbe ed knov ses the c nter as m . (examp	assical of s the nto k r k of vn a losest edoid le →			<i>In front</i> using th <i>D</i> (fig. 2 time seri	, k-medoids with (k=4) e dissimilarity scores in (k=4) chooses the 4 annual es as centers (medoids).	$2 \qquad M \qquad 1$ $1 \qquad 0 \qquad 1884 \qquad M$ $1 \qquad 1931 \qquad 1958 \qquad 955 \qquad 1995 \qquad 1995 \qquad 1995 \qquad 1995 \qquad 1995 \qquad 100 $	M. M	M. M	
Label clusters based on a deep analysis of its medoids and the characteristics it shares with the cluster.	medoids	RR_{max} $[mm. h^{-1}]$	σ_R $[mm. h^{-1}]$	H _d [mm]	S _R [jour]	pic max	Nb Events	with extreme variability) 4.0 4.5 4.0 4.5 4.0 Classe représentée par 19 6.0 Classe représentée par 19	 4.5 4.0 Classe représentée par 1953 Classe représentée par 1995 Classe représentée par 1931 Classe représentée par 1884 Classe représentée par 1884 Classe représentée par 1884 Classe représentée par 1884 	 Classe représentée par 1953 Classe représentée par 1995 Classe représentée par 1931 Classe représentée par 1884 0.30 		
	1884	2.04	0.22	459.90	142	June 28 th	56	រុី 3.0 មុខ 2.5	تا 3.0 ج ق 2.5	1.5 1.0 0.5 1.0 1.5 1.0 0.5 1.0 1.5 1.0 0.5 1.0 1.5 1.0 0.5 1.0 1.5 1.0 0.5 1.0 0.5 1.0 0.5 0.25 0.20 0.15 0.15 0.5 0.25 0.20 0.15 0.15 0.5 0.15 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.		
	1931	1.04	0.16	457.40	149	Oct. 23 rd	60	2.0 2.0 1.5	2.0 2 1.5			
	1953	0.53	0.13	382.90	146	Aug. 24 th	53	1.0	1.0		0.15 0.15 0.5 1.0 1.5 2.0 2.5 3.0 3.5 4.0 4.5	
	1995	0.66	0.14	378.10	134	July 4 th	62	C1 (drought years)	250 300 350 ximal [<i>jour</i>]	400 500 600 700 800 900 R _d [mm]	$R_{max}[mm.h^{-1}]$	

- Considering the appearance frequency of each cluster in seven successive 26 years' time intervals between 1873 and 2019 allows studying the possible evolution of precipitation.

Cluster Temporal window C2 **C1**

Analyze the evolution of clusters frequencies over time.

Table 2. shows The temporal evolution of frequencies for each cluster [% by column]. \rightarrow

- \Rightarrow With a coarse time window:
- the presence of the first cluster C1 (drought years) decreases over time;
- while that of C4 (remarkable years with extreme variability) increases or

stagnates at the same time

END

which corroborates the assumption that the precipitation variability increases over time (# warmer climate) between 1873 and 2019.

IV. CURRENT WORK

The proposed method shows that :

- there exists a temporal evolution of precipitation when studiying Paris-Montsouris time series. Limitations of the current work include :
- the fact that we have not yet demonstrated the robustness of the results in table 1, i.e. test if other rainfall time series measured in Paris area could lead to similar results.

Future work :

- Investigate if results show a real impact of climate change rather than just representing a temporal correlation.
- exploring the correlation between the defined clusters and environmental parameters (such as temperature).
- We also plan to extend the study area and to use other rainfall time series.

▲					
1873	1899	29	19	16	07
1892	1918	21	19	19	14
1911	1937	18	15	29	11
1930	1956	18	11	24	25
1949	1975	21	23	81	21
1968	1994	18	23	11	21
1987	2018	09	28	16	36
Grad	lient	7	\mathbf{V}	フレ	7

V. REFERENCES

- Pendergrass, A. G., Knutti, R., Lehner, F., Deser, C. and Sanderson, B. M., "Precipitation variability increases in a warmer climate", Scientific reports (2017).
- Verrier S., & al., Multiscaling properties of rain in the time domain, taking into account rain support biases, Journal of Geophysical Research: Atmospheres, American Geophysical Union, (2011)
- Akrour N., & al., Simulation of yearly rainfall time series at microscale resolution with actual properties: Intermittency, scale invariance, and rainfall distribution Water Resources Research, American Geophysical Union, (2015).
- Dilmi, M. D., Barthès, L., Mallet, C., Chazottes, A., "Iterative multiscale dynamic time wapring : a tool for rainfall time series comparison". International Journal of Data Science and analytics, (2019).
- Sakoe, H. and Chiba, S., "A similarity evaluation of speech patterns by dynamic programming", Dig. Nat. Meeting, Inst. Electron. Comm. Eng. Japan, (1970).
- Kaufman, L., and Rousseeuw, P. J., "Clustering by means of medoids", in In : Dodge Y & editor, ed., , North Holland/ Elsevier, p 405-416, (1987).