Earth Sciences Department



Supercomputing Centro Nacional de Supercomputación

WGSIP25's Centre Update:

BSC

Ángel G. Muñoz Head, Climate Services Team (CST) Ramón y Cajal Fellow

Earth System Services (ESS) Group

Grant RYC2021-034691-I, funded by MCIN/AEI/10.13039/501100011033 and the European Union NextGenerationEU/PRTR

Earth Sciences Department

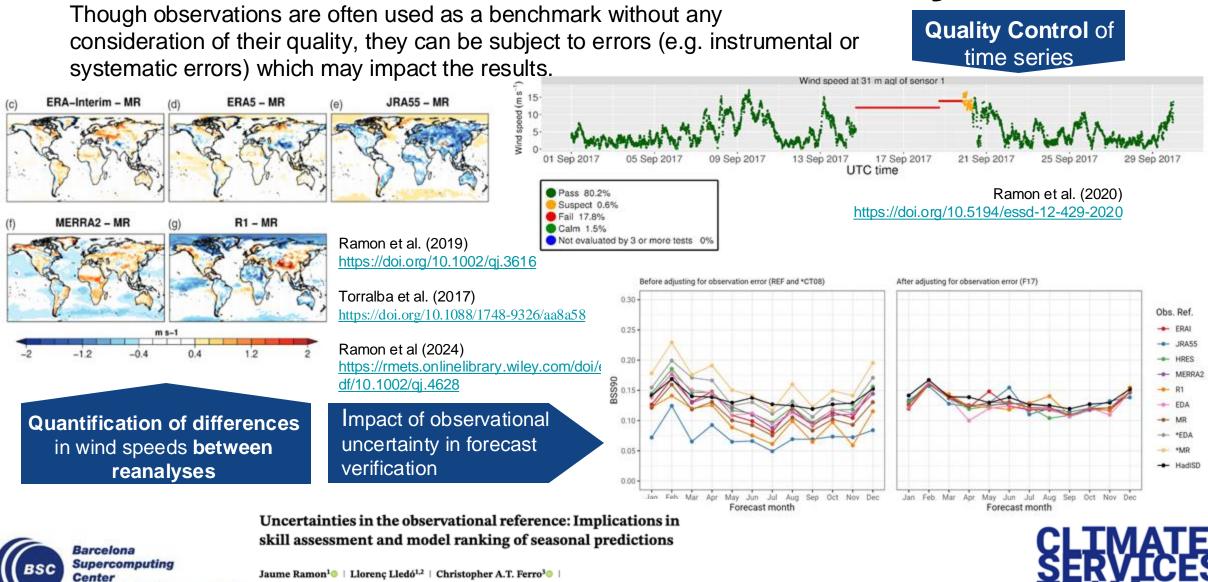
Environmental modelling and forecasting using process-based and artificial intelligence models, with a particular focus on weather, climate and air quality. This includes transferring solutions to support the main societal environmental challenges through data applications



Climate services across sectors

	Sources	Description	С	ompetences		Target	
	PAST OBSERVATIONS	Includes station data, gridded datasets or reanalysis	Tall towers data Observational				
scale	SUB-SEASONAL FORECASTS	Fill the gap between weather and seasonal forecast			Use cases	APPLICATIONS Wind Energy	
Temporal scale	SEASONAL FORECASTS	Takes advantage of slowly changing parts of the climate system	Assessment and	Indicators	and operational services	Agriculture Retail Health	
Ţ	DECADAL FORECASTS	Fill the gap between seasonal and long- term projections	verification			Herding Urban areas	
	CLIMATE PROJECTIONS	Explore long-term climate change impacts			Seamless climate projections		
	CUS-AFRICA VITIGED			Ate Prediction system		PIISA	arth EUROPEAL CHINE FOR
Climate Services for Clean Energy	Erise	ArcticXchange	ibalert Infectious Disease decision-support tools and Alert systems		0	Al for Drought	
Bsc Barcelona Supercomput Center Centro Nacional de	ing MED-GO	D	A	SPECT	IMPE 4CHA		ERVÍCES EAM

Observational uncertainty



Centro Nacional de Supercomputación Francisco J. Doblas-Reyes^{1,4}

Bias adjustment and calibration

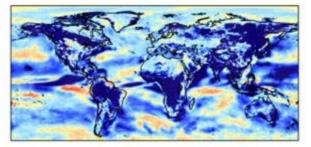
relative

Observed

0.1

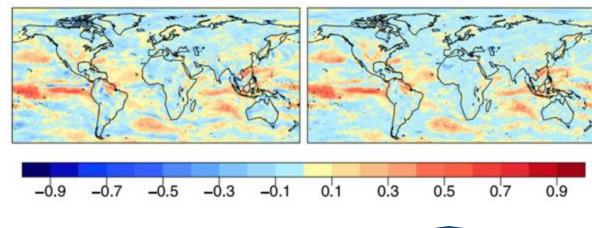
Through bias adjustment, we remove systematic errors and improve the reliability of climate predictions.

(a) Uncorrected



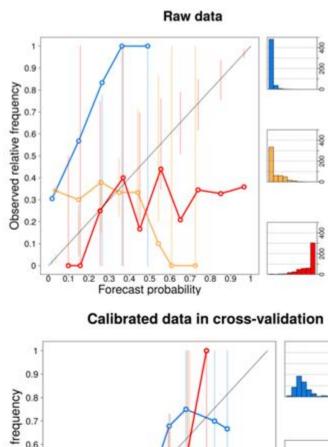
(b) Simple bias corrected

(c) Calibrated





Fair RPSS for tercile events of winter 10-m wind speed forecasts



0.3 0.4 0.5 0.6 0.7 0.8 0.9 Forecast probability



Torralba et al. (2017) https://doi.org/10.1175/JAMC-D-16-0204.1

> Reliability diagram exemplifying the difference between raw and calibrated seasonal predictions

Calibration function from the <u>CSTools</u> package

Pérez-Zanón et al. (2022) https://doi.org/10.5194/gmd-15-6115-2022



Forecast quality assessment

The forecast quality assessment (or forecast verification) is a necessary step to evaluate the performance of the predictions and should be provided with every forecast product.

Analysis of probabilistic and deterministic metrics are performed. Given the probabilistic nature of the climate forecasts, **scores** and **skill scores** are commonly used.



Scorecards: Visualisation of verification metrics to measure the performance of models and estimate the reliability of climate predictions

Doblas-Reyes et al. (in preparation)

Barcelona Supercomputing Center Centro Nacional de Supercomputación NO SON

FairRPSS for SPEI6 predictions of September (ref. Period 1985-2017), using hindcast data

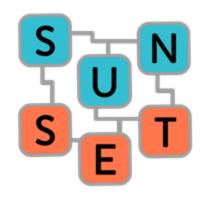
Torralba et al. (2017) https://doi.org/10.1175/JAMC-D-16-0204.1 Delgado et al. (2022) https://doi.org/10.1175/JCLI-D-21-0811.1 Solaraju-Murali et al. (2019) https://doi.org/10.1088/1748-9326/ab5043

m temperat	ture of	ECM	w	FS	yst	em	5		1	ow						1	0E						3	OE																									
ef: ERA5 1993-2	(3.6)					Me	an	Bia	as ((K)									C	orre	lat	ion										CR	PSS	8										RP:	55				
1. ERAD 1999-4	0101						Sta	rt di	ate											Star	t dat											Star	t date										8	Start	date				
Region	Forecast	Jan	feb	Mar	Ap	May	y po	n ji	a a	og 1	iep (Oct 1	Nov	Dec	jan	feb	Mar	Apr	May	Jun	Jut	Aug	i Sep	Oct	Nov	e Dec	Jan	feb	Mar	Apr	Мау	Jun	Jul	Aug	Sép	Oct	Nov	Dec	Jan	Feb	Mar	Apr	May	jue	jul 4	hog 5	iep d	bet N	iov D
	1	4.39	4.0	4.04	4.0	4.13	E -44	0 A	38. 4		e.34	4.17	4.81	421		649	***	***	-			***	-				-8.13	4.05	4.05	-8.11	4.13	4.0	- 4.80	-8.49	4.07	-434	8.07	4.24	5.48	830	***	-	6.48				-0.0		
	2	4.25	4.30	-0.36	4.5	- 4.13	- 44	4. 4	20 -4	- 36		4.28	4.22	4.24		441								4.14			4.86	4.34	-8.34	-8.33	*.**	4.56	-842	-9.62	434	4.51	**	-6.15	9.47	8.88	4.11	833	6.25	6.25	121	-	a) 1	40 B	36 a
Tropics	3	-8.21	4.18	-8.81	4.0	4.1	4.43	a 4	M -0			4.11	4.20	4.24	-	-	8.51	8.51	1.18	8.36	8.58	-	8.31	4.81	8.54	-	-10	440	- 14	438		4.72	476	4.68	4.63	4.34	44	434	6.25	8.23	4.0	4.00	833	-	an j	823 F	-	an e	-
(305-30N)	4	4.21	4.15	4.0	4.0	- 4.8	-43	H -4	17 A	ur e	8.38	4.59	8.28	421	2.50	4,88	8.81	-	-8.54	8.84	-	- 11	0.54	8.01	8.88		**		-15	4.73		-0.02	4.16	-0.68	.4.40	4.51	-8.81	-848	8.24	834	8.17		0,25	8.17		8.58 B		.18	-
	5	.8.10	-8.13	-6.00	.4.4	4.1	-43	17 -4.	n 4		a.38	4.85	4.85	4.24		4.96		4.54	6.52	-	4.94	6.34		4.50	=.55	8.54		4.98	-4.76	-8.87	4.82	-4.83	-8,16	-8,48	4.95		-8.87	4.88	9.48	9.54	8.15	8.5.1	8.15	0.12	8.25	837 F			
	6	4.17	4.13	-8.08	-4.8	-415		20 -4.	28 .4	1.00	1.18	-	4.19	-6.21		8.86	4.10		-8.84		8.44	4,64	8.54	6.16	8.34	6.45	-8.65	4.76	4,92	-4.11	*8	4.79	-8.71	-0.98	4.50	4.33	4.56	-639	0.11	8.13	8.11	833	6,56	8.15	8.3.5	817 B		34 8	10 4
	1	4.0	4.60	8.58	0.34		-44		88 I I I			8.47		44	-	888	817	4.74	8.84	8.85	-			-			-8.60		- 449	44		-6.81	436	.8.34	4.94	8.68	8.83	8.82	842						441	1.24 F	a 9	-	28 A
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ctra-tropical NH		6.62	633	6.87	4.0	-43	1 44	6 A.	12	•	-	441		431	627	629	4.25	8.07	633	*,8	4.38	4.31	8.34	8.26	. 6.33	. 8.30	4.78	438	-16	-8.34	+7	-4.0	410	4.0	436	4.30	436	-4,38	6.03	8.04	6.03	8.62	6.87	4.87	6.87 Y				
(30N-90N)	4	8.84	8.08	4.36			-43	13 44			1.17		4.38	-6.81	8.28	8.28	*22	8.21	828	8.83	8.38	*20	8.25	8.28	8.27	8.33	-4.0	-	-4.40	4.74	4.92	-8.87	-8.10	8.38	:4.38	-8.38	4.11	-8.85	0.00	8.82	8.85	4.81	0.88	***		4.52 4	in a		
	5	4.53		-8.81		42		•		48 4		4.30	4.11	++1	8.29	6.81	411	8.54	8.37		4.24	4.0	6.25	8,0	4.23	4.17	***	-0.62	4.78	-0.64	4.17	6.25	-	-	4.34	411	430	-644	6.88	8.05		-4.45	-	4.05	4.60				1838
	6	4.44	4.62	443	4.1	8.81	8.1	3 6.	19 -4	LOR -		4.62	9.06	-6.82	8.28	6.25	8.23	8.29	6.38	8.27	6.25	4.31	8.25	8.17	8.38	8.17	4.6	4.79	447	4.43	-8.54	-4.36	411	40	4.35	4.86	-8.86	444	0.00	8.63	8.80	8.83	0.05	4.82	8.62	8.82 E	1.04 0		1.09 -6
	1	6.88	4.44	4.00	8.21	. 8.20	-44		24 ×	10.1	-10	6.88 ()	4.06	444	-	6.96	-	-									***	4.88	4.03	8.85	0.83	8.80	0.63	8.85	8.45	8.88	8.85	-638	6.38	8.01	4.31	-	639	840	111 I	627 T		-	-
	2	8.49		8.36	8.27	-44	- 44	• •	17 4	36 - 1	416	438	4.35	8.06	1.39	8.H	8.84	8.23	8.17	8.25	6.28	. 433	- 8,34	8.96	8,33		4.16	4.43	-8.18	4.11	1836	-833	433	18.34	- 8.27	4.31	4.65	-141	0.38	8.08	0.85	8.80	0,69	8.83			106 1	47 . 8	45 8
rtra-tropical SH	3	***	-	8.40		8.34	83		24 8	n -	448	4.28	441	1.11	4.88	6.33		8.14	6.18		6.28	8.24	8.0	8.25	8.33	8.88	-	4.37	4.16	4.94	4.07	-	838	438	438	4.54	4.75	-0.94	0.24	8.05	6.82	-6.61			-				
(305-905)	4	1.36	-	6.13	8.34	-4.0			-	11	1.11	4.02	-	1.00	8.84	4.24	8.18	8.58	8.13	4.13	8.24	6.27	8.81	8.28	8.28	8.29		4,38	-	-	8.0	4.0	-6.07	-8.00	4.42	4.87	1.00	-3.49	9.85	8.61	8.81	-4.62	-4.65	-0.66			-	-	
	5		8.87		8.00	-4.8			** *	41	8.87	6.85	1.00	1.05	8.28	6.58	8.1.8		6.39	8.25			8.21	6.21	8,37		4.56	4.30	-4.00	4.8	-	-444	***	-8.45	-1.00	-1.00	-1.00	-8.81		-	-8.62			-4.45	-			-	
	6	8.25	8.58	0.64	8.85	40	1	1	38 4	.00	-	100	1.25	1.12	8.18	0.28	8.19	8.26	8.25		8.21	4.20	8.27	8.28	8.21	8.36			4.10	-0.32	-	0.40	-8.67	-1.00	13.00	-1.00	4.00	4.70	0.00	8.00	4.80	8.98	0.63	8.94	6.43			-	- 68.

FairRPSS. Target month:09. SD:05.

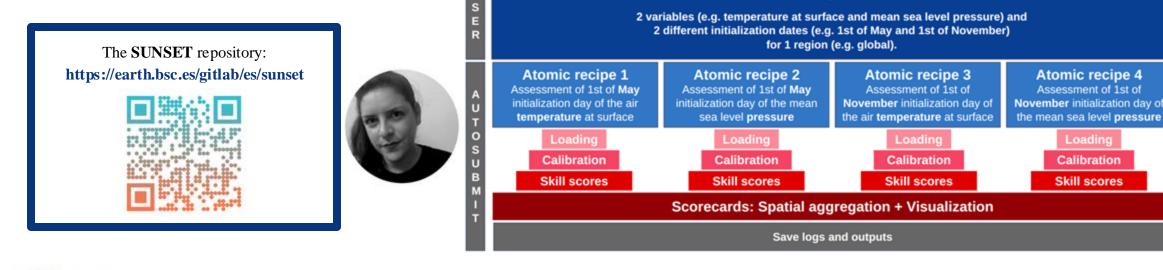
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SUNSET: the SUbseasoNal-to-decadal climate forecast post-processiNg and asSEssmenT suite



SUNSET is a software suite developed collaboratively at the Barcelona Supercomputing Center, that aims post-process climate forecast outputs to provide climate services for sub-seasonal, seasonal and decadal time scales.

Its **modular design** allows the technicians and researchers flexibility in defining the required products, post-processing steps, outputs and visualization options. These computations can then be easily parallelized on HPC platforms, making use of the Autosubmit workflow manager.





Núria Pérez-Zanón, Victòria Agudetse, Lluis Palma, Carlos Delgado-Torres, Nadia Milders, Eren Duzenli, Alba Llabrés-Brustenga, Bruno de Paula Kinoshita, Pierre-Antoine Bretonnière, Albert Puiggròs, Ángel G. Muñoz

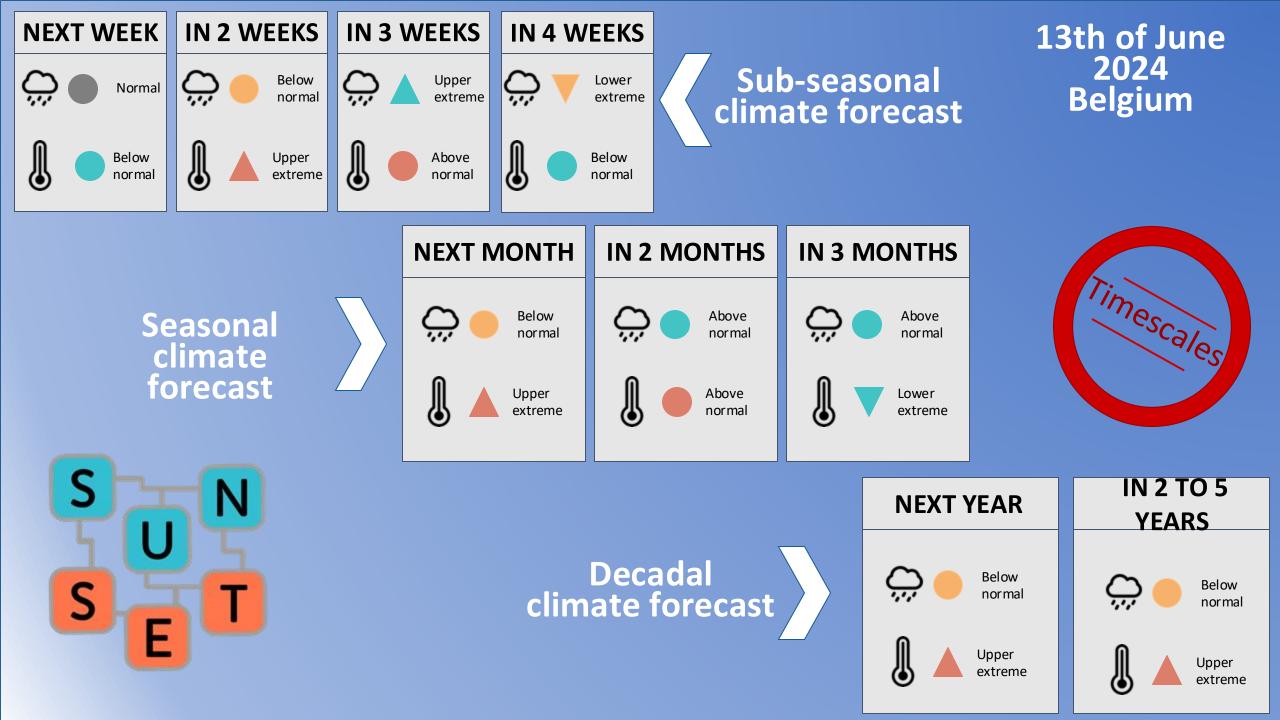
Loading

Calibration

Skill scores

RECIPE

Pérez-Zanón et al., (2022, 2024); Pérez-Zanón et al., in



Exploring downscaling methods across timescales

The R package **CSDownscale**

downscaling climate data, either to

contains seven methods for

Providing climate predictions with high resolution is very much needed for some applications, and here is where statistical downscaling comes to help.

a finer grid or to a point location. Intbc / CST_Intbc • Quantile mapping Coarse scale Bias-Simple bias correction model data EVMOS(variance inflation) correction • mse min • crps min Interpolation rpc based no need for training data (observations are not used Intlr / CST Intlr Linear here) • Basic regression Interpolation / • Large-scale **CST_Interpolation** • 9nn Bilinear Bicubic LogisticReg / Conservative **CST** LogisticReg Logistic Second order conservative • Ens mean Fine scale Inverse distance weighting regression Ens mean sd model data Large area fraction • Sorted members • Nearest neighbour regrid or point location Barcelona Analogs / Supercomputing Analogs **CST** Analogs Center antro Nacional de Supercomputación Duzenli et al., sub judice

Comparing Constraining Methods in NTCP

[©]Near-Term Mediterranean Summer Temperature Climate Projections: A Comparison of Constraining Methods[®]

PEP COS[®], RAÜL MARCOS-MATAMOROS[®], MARKUS DONAT[®], a.c. RASHED MAHMOOD, ^{a,d} AND FRANCISCO J. DOBLAS-REYES^{a,c}

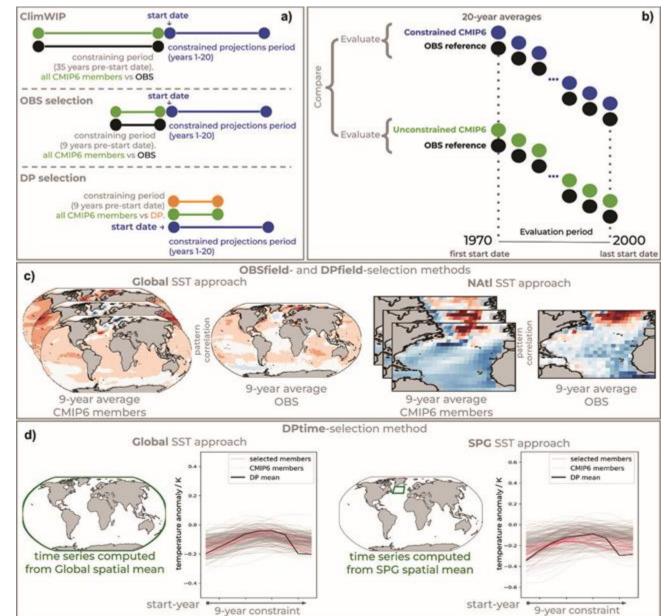
^a Earth Sciences Department, Barcelona Supercomputing Center, Barcelona, Spain
^b Department of Applied Physics, University of Barcelona, Barcelona, Spain
^c Institució Catalana de Recerca i Estudis Avançats, Barcelona, Spain
^d Danish Meteorological Institute, Copenhagen, Denmark

Fig. 1. Diagrams illustrating the main features of the constraining methodologies. (a) Constraining periods, start dates, and constrained projection diagrams for the ClimWIP, OBS, and DP selection methodologies (note that DPfield and DPtime use the same period for the constraint). (b) Process for the evaluation of the constrained and full CMIP6 ensembles against observational reference. (c) The OBSfield- and DPfield-selection methods can compute the SST pattern correlations over different target regions, which in this paper are either global (Glob) or the North Atlantic (NAtl). (d) The DPtime approach computes the mean absolute error between the reference time series and the rest of the CMIP6 ensemble, which are computed from either the Glob or the North Atlantic SPG area-averaged SSTs.

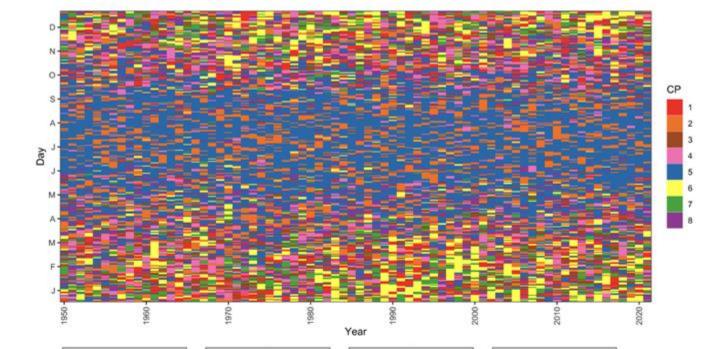




Cos et al., 2024. JClim



Flow-dependent, cross-timescale physical diagnostics



1 NOVEMBER 2024

OLMO ET AL.

Cross-Time-Scale Analysis of Year-Round Atmospheric Circulation Patterns and Their Impacts on Rainfall and Temperatures in the Iberian Peninsula

MATÍAS OLMO©,^a PEP COS,^a ÁNGEL G. MUÑOZ,^a VICENT ALTAVA-ORTIZ,^b ANTONI BARRERA-ESCODA,^b DIEGO CAMPOS,^a ALBERT SORET,^a AND FRANCISCO DOBLAS-REYES^a

> ^a Barcelona Supercomputing Center, Barcelona, Spain ^b Meteorological Service of Catalonia, Barcelona, Spain



Olmo et al., 2024. JClim

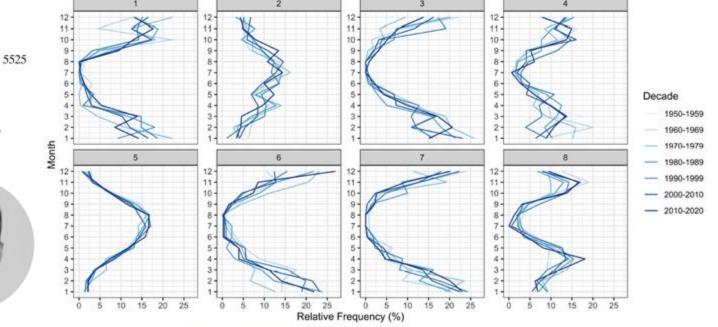


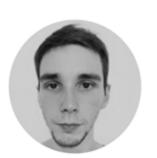
FIG. 7. (top) Klee diagrams (Muñoz et al. 2016) for the D2.8 classification. The x axis presents the years (1950–2022) and the y axis presents each day of the year. (bottom) The monthly frequencies of the CPs separated by decades (blue colors).

Flow-dependent, cross-timescale model diagnostics

Filtering CMIP6 models in the Euro-Mediterranean based on a circulation patterns approach

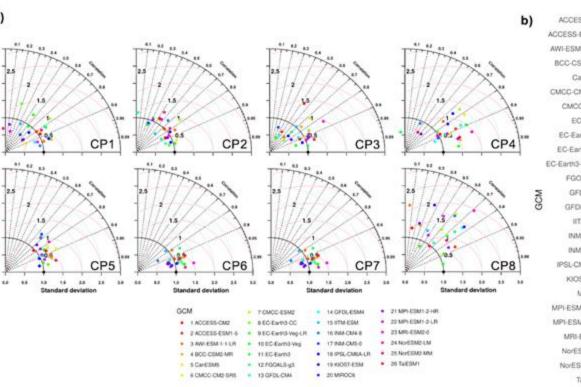
Olmo ME¹*, Cos P¹, Campos D¹, Muñoz ÁG¹, Altava-Ortiz V², Barrera-Escoda A², Jury M^{1,3}, Loosveldt-Tomas S¹, Bretonniere PA¹, Doblas-Reyes F^{1,4}, Soret A¹.

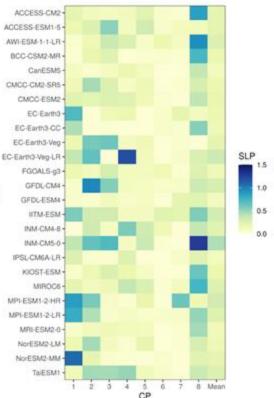
a)



Olmo et al., sub judice







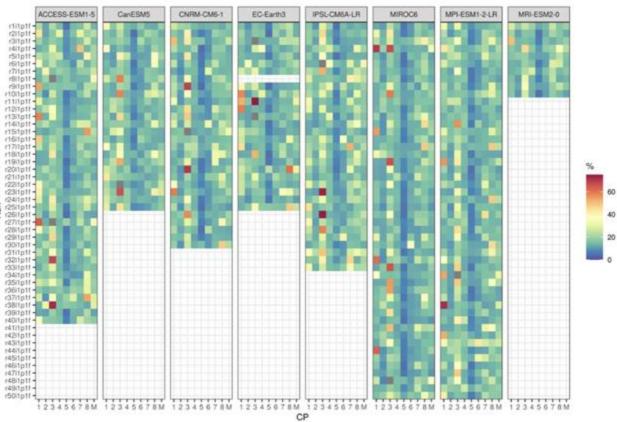
Representation of internal variability by models is key

On the role of model internal variability in reproducing synoptic circulation

patterns in the Euro-Mediterranean

Matías Olmo^a, Pep Cos^a, Diego Campos^a, Ángel G. Muñoz^a, Margarida Samso^a, Albert Soret^a, Francisco Doblas-Reyes^a

^a Barcelona Supercomputing Center, Barcelona, Spain

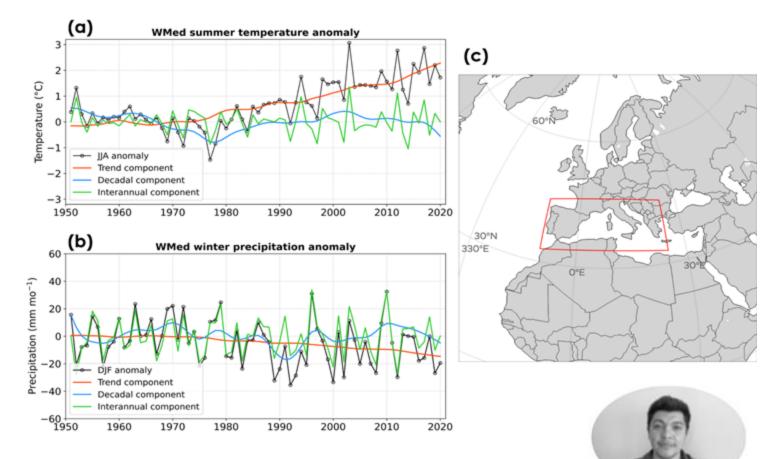






Muñoz et al (2017) Olmo et al., sub judice FIG. 4. ErrorCycle metric for each GCM run, expressed as a percentage of days with respect to the climatological frequency in the ERA5 reference. Results are presented for every CP, separately, and for the weighted mean of all CPs (M).

Diagnosing Observed and Modelled Trends via Timescale Decomposition



Campos et al., sub judice

Figure 1. Timescale decomposition of the Western Mediterranean (a) winter precipitation anomalies, and (b) summer temperature anomalies. Long-term trends (red), decadal (blue), and interannual components (green) are shown. (c) Map of the area of study. The gray area (land-only) within the red box represents the defined Western Mediterranean (WMed).

Regional aspects of the recent observed trends in the Western Mediterranean: Insights from a Timescale Decomposition Analysis

D. A. Campos^{1,2}, M. E. Olmo¹, P. Cos^{1,2}, F. J. Doblas-Reyes^{1,3}, and Á. G. Muñoz¹

¹ Earth Sciences Department, Barcelona Supercomputing Center, Barcelona, Spain.
² Facultat de Física, Universitat de Barcelona, Barcelona, Spain.
³ Catalan Institution for Research and Advanced Studies, ICREA.



Causal ID of Sources of Predictability

LK causality between (snow, z500 and sst), and t2m over the box

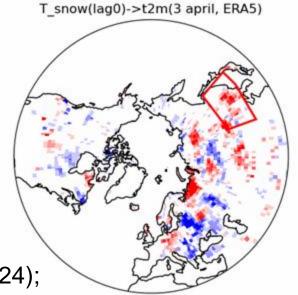
(colors indicate statistically significant, p<0.05)

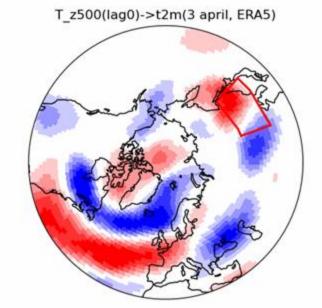
Ardilouze, Materia and Muñoz (2024); Muñoz et al. (2023); Ardilouze, Muñoz, Materia (in prep) Muñoz (2024; in prep)

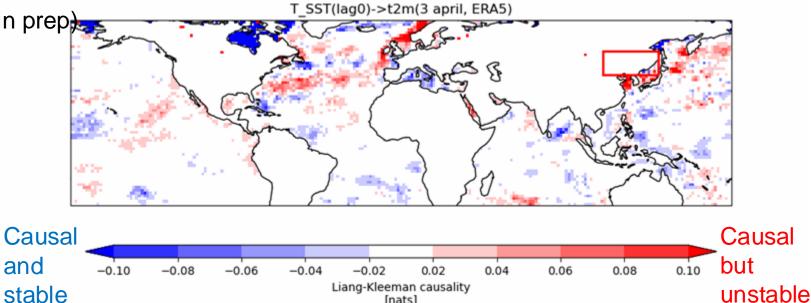
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[nats]

Grant RYC2021-034691-I, funded by MCIN/AEI/10.13039/501100011033

and the European Union NextGenerationEU/PRTR

Earth Sciences Department



Supercomputing Centro Nacional de Supercomputación

WGSIP25's Centre Update:

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