







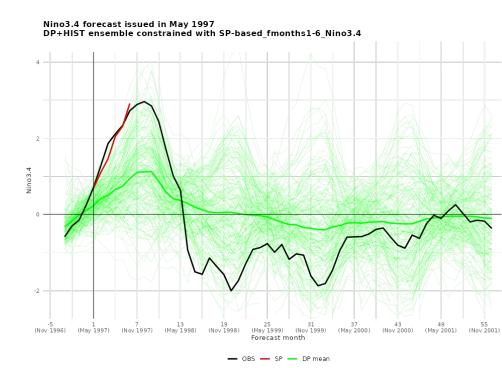
Seamless climate information for the next few months to multiple years: merging of seasonal and decadal predictions, and their comparison to multi-annual predictions

Carlos Delgado-Torres, Markus G. Donat, Núria Pérez-Zanón, Verónica Torralba, Roberto Bilbao, Pierre-Antoine Bretonnière, Margarida Samsó-Cabré, Albert Soret, and Francisco J. Doblas-Reyes

carlos.delgado@bsc.es

Motivation

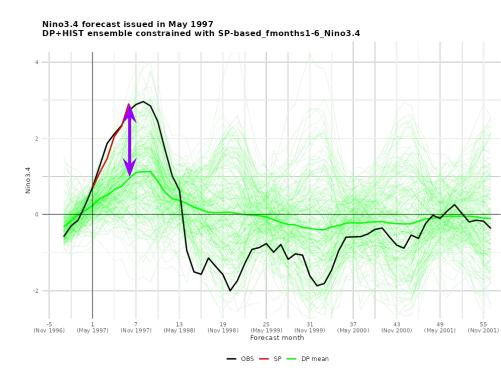
- Need for climate information that spans multiple timescales. Shifting from seasonal forecasts (used during the first few months) to decadal forecasts (for subsequent years) introduces inconsistencies
- By combining seasonal and decadal forecast, we could:
 - Improve forecast quality (by transferring short-term predictability into longer time scales)
 - Increase forecast consistency at seasonal and decadal timescales (by avoiding "jumps" when switching from one horizon to the other)





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- Need for climate information that spans multiple timescales. Shifting from seasonal forecasts (used during the first few months) to decadal forecasts (for subsequent years) introduces inconsistencies
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Data

Climate predictions

Time scale	Forecast months	Total members	N	Forecast system	Ensemble members
Seasonal	1-6	25	1	SEAS5	1-25
Multiannual	1-24	70	1	CMCC-CM2-SR5-i5	r1-10i5p1f1
			2	EC-Earth3-i5	r1-20i5p1f1
			3	EC-Earth3-HR	r1-15i1p1f1
			4	ECMWF	1-25
	1-60	197	1	BCC-CSM2-MR	r1-8i1p1f1
Decadal			2	CanESM5	r1-20i1p2f1
			3	CESM1-1-CAM5-CMIP5	1-40
			4	CMCC-CM2-SR5-i1	r1-10i1p1f1
			5	CNRM-ESM2-1	r1-10i1p1f2
			6	EC-Earth3-i1	r1-10i1p1f1
			7	EC-Earth3-i2	r6-10i2p1f1
			8	EC-Earth3-i4	r1-10i4p1f1
			9	FGOALS-f3-L	r1-3i1p1f1
			10	HadGEM3-GC3.1-MM	r1-10i1p1f2
			11	IPSL-CM6A-LR	r1-10i1p1f1
			12	MIROC6	r1-10i1p1f1
			13	MPI-ESM1-2-HR	r6-10i1p1f1
			14	MPI-ESM1-2-LR	r1-16i1p1f1
			15	MRI-ESM2-0	r1-10i1p1f1
			16	NorCPM1-i1	r1-10i1p1f1
			17	NorCPM1-i2	r1-10i2p1f1



Data

Climate projections (hist+ssp245)

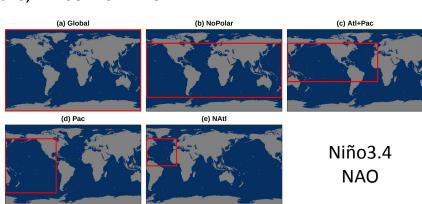
Total members	N	Climate model	Ensemble members	N	Climate model	Ensemble members
	1	ACCESS-CM2	r1-5i1p1f1	17	FGOALS-g3	r4i1p1f1
264	2	ACCESS-ESM1-5	r1-12i1p1f1	18	FIO-ESM-2-0	r1-3i1p1f1
	3	BCC-CSM2-MR	rlilp1f1	19	GISS-E2-1-G	r1-10i1p1f2
	4	CAMS-CSM1-0	r1-2i1p1f1	20	HadGEM3-GC3.1-LL	rlilp1f3
	5	CAS-ESM2-0	r(1,3)i1p1f1	21	INM-CM5-0	rlilp1f1
	6	CESM1-1-CAM5-CMIP5	1-40	22	IPSL-CM6A-LR	r(1:-6,10-11,14,22,25) ilp1f1
	7	CESM2-WACCM	r1-3i1p1f1	23	KIOST-ESM	rlilp1f1
	8	CMCC-CM2-SR5	rlilp1f1	24	MIROC-ES2L	r1-30i1p1f2
	9	CMCC-ESM2	rlilp1f1	25	MIROC6	r1-50i1p1f1
	10	CNRM-CM6-1-HR	rlilp1f2	26	MPI-ESM1-2-HR	r1-2i1p1f1
	11	CNRM-CM6-1	r1-6i1p1f2	27	MPI-ESM1-2-LR	r1-10i1p1f1
	12	CNRM-ESM2-1	r1-10i1p1f2	28	MRI-ESM2-0	r1-5ilp1f1
	13	CanESM5-CanOE	r1-3i1p2f1	29	NESM3	r1-2i1p1f1
	14	CanESM5	r1-25i1p2f1	30	NorESM2-LM	r1-3i1p1f1
	15	EC-Earth3	r(2,7,10,12,14,16-25) ilp1f1	31	NorESM2-MM	rlilp1f1
	16	FGOALS-f3-L	rli1p1f1	32	UKESM1-0-LL	r(1-4,8)i1p1f2



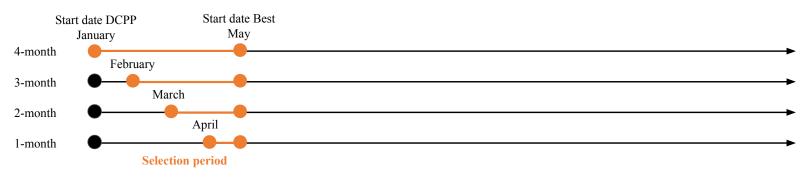
Methods

- Selecting the best 30 members from
 - decadal predictions (197 members from 17 models)
 - climate projections (264 members from 32 models)
- The selection is based on
 - o recent observations (1-4 months)
 - seasonal predictions (SEAS5; forecast months 1-6)
- It can be **updated at any time of the year** (with updated observations or seasonal predictions)
- All members are bias-adjusted to have the correct mean and variance
- Tested different constraints based on
 - variables: sea surface temperature, sea level pressure, Niño3.4 or NAO
 - o metrics: ACC, centered-RMSE, uncentered-RMSE
 - o regions: Global, NoPolar, Atl+Pac, Pac, NAtl
- With all combinations:
 - 128 OBS-based constraints
 - 192 SP-based constraints

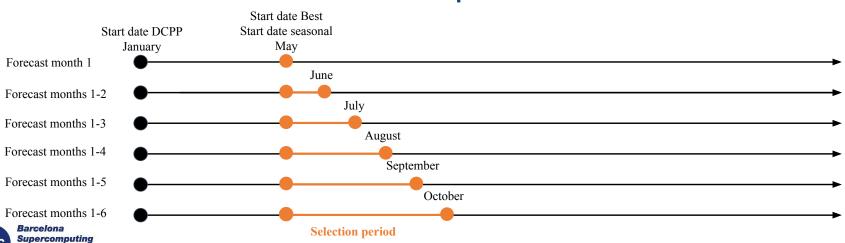




Based on recent observations



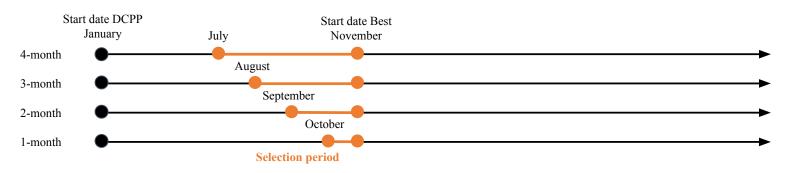
Based on seasonal predictions



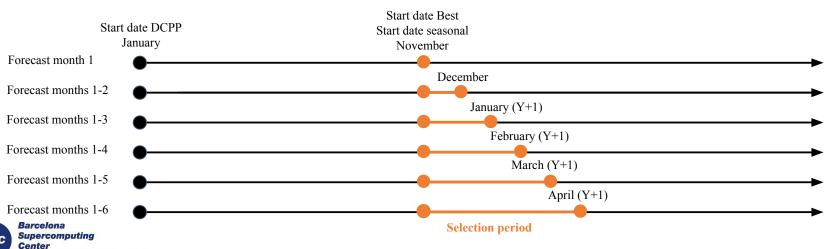
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Based on recent observations

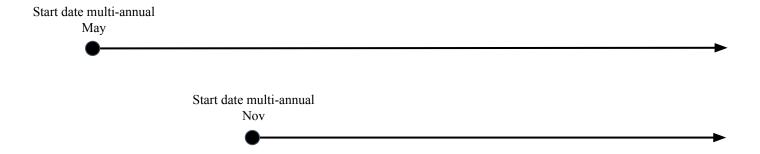


Based on seasonal predictions



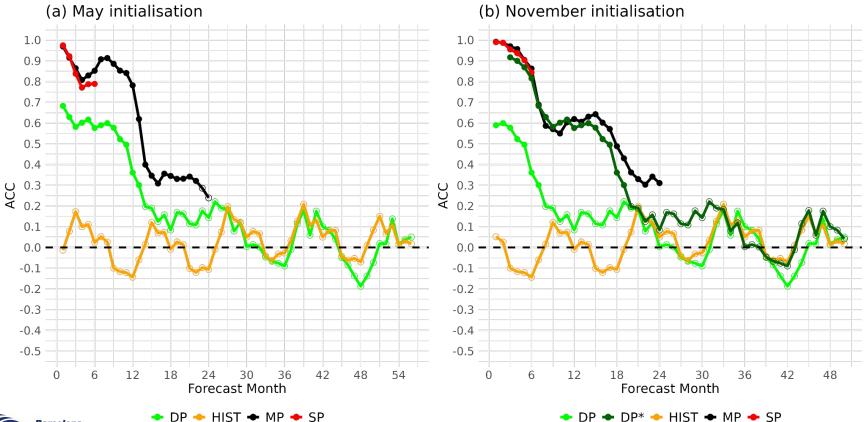
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Initialised multi-annual predictions



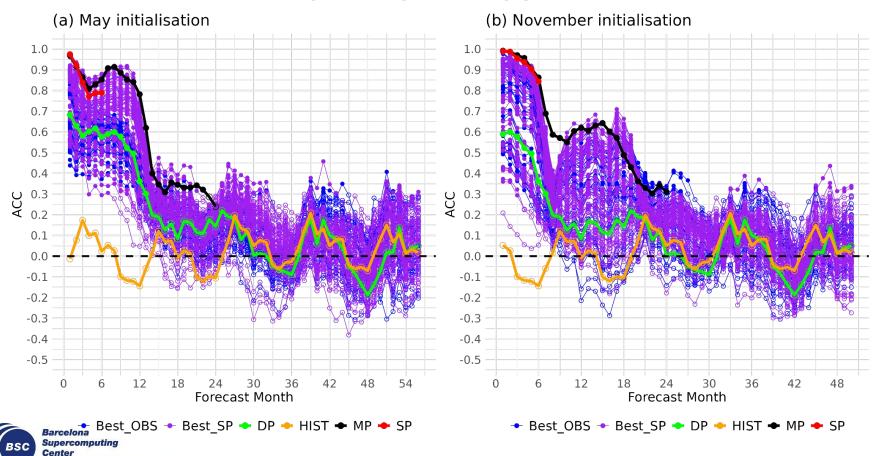


Skill for Niño3.4



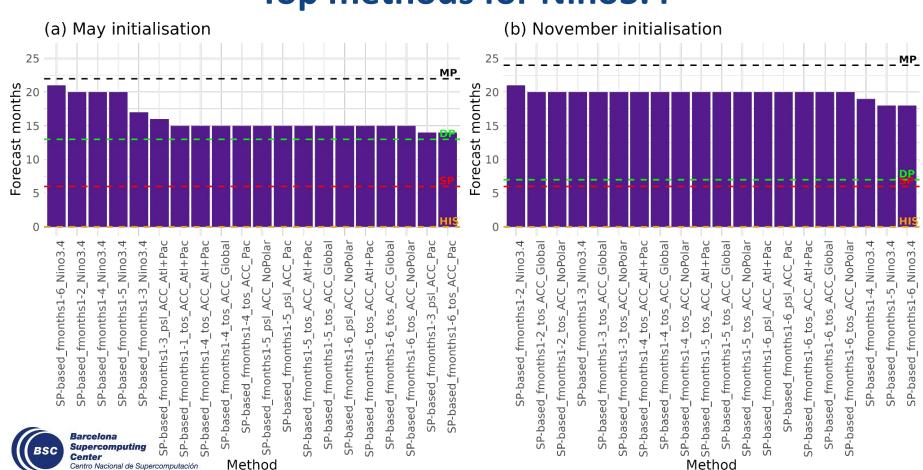


Skill for Niño3.4

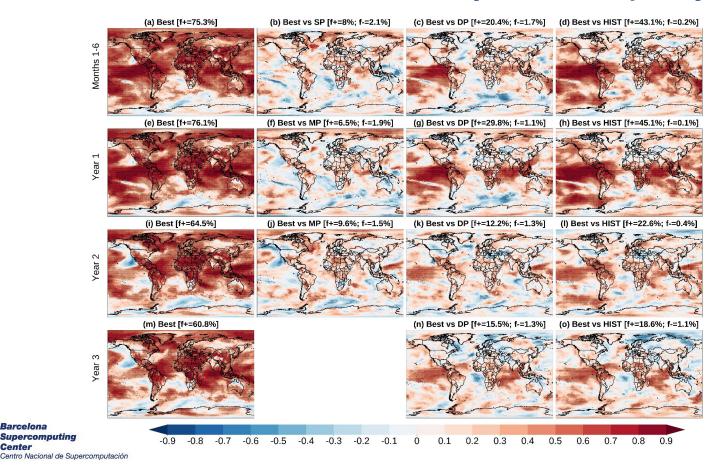


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Top methods for Niño3.4

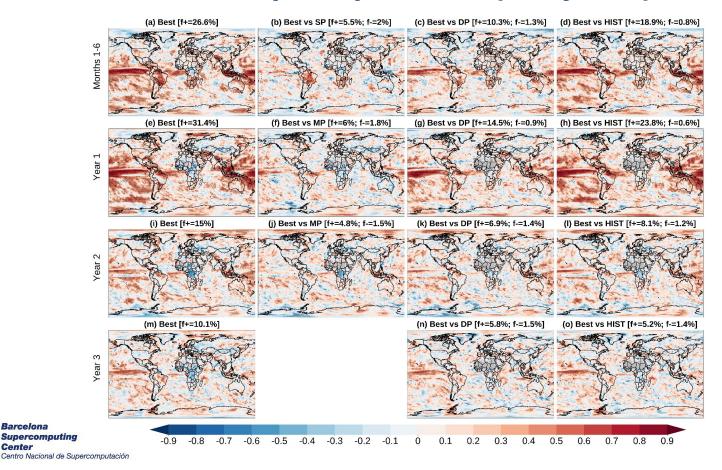


Skill for near-surface air temperature (May init)



Barcelona

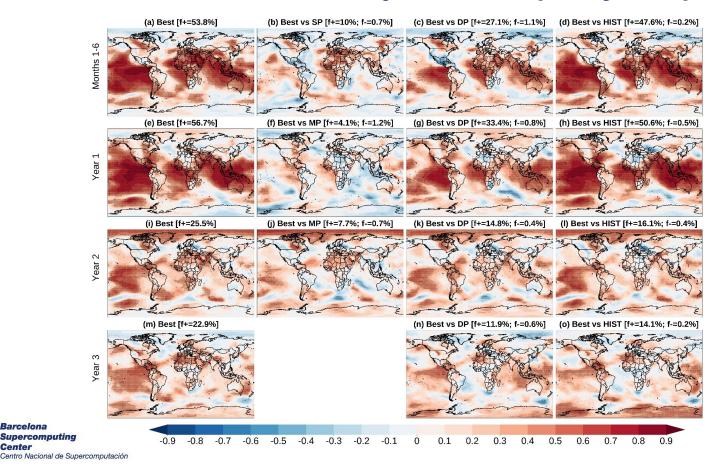
Skill for precipitation (May init)



Barcelona

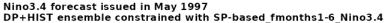
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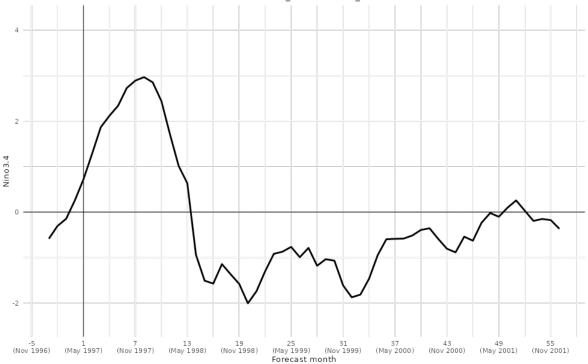
Skill for sea level pressure (May init)



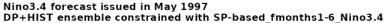
Barcelona

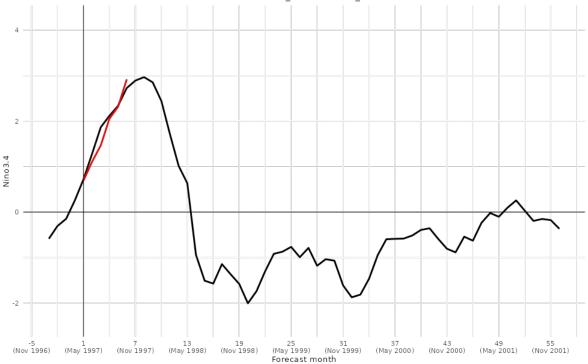
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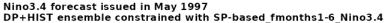


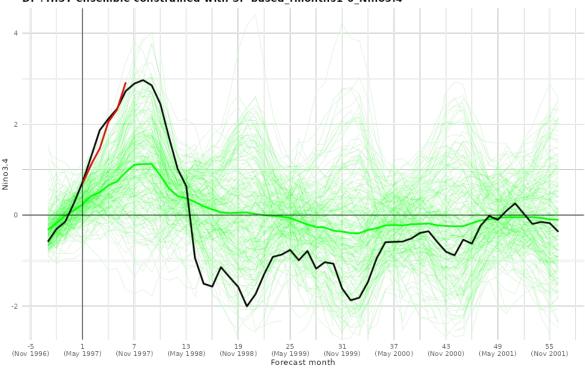




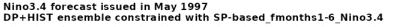
— OBS — SP

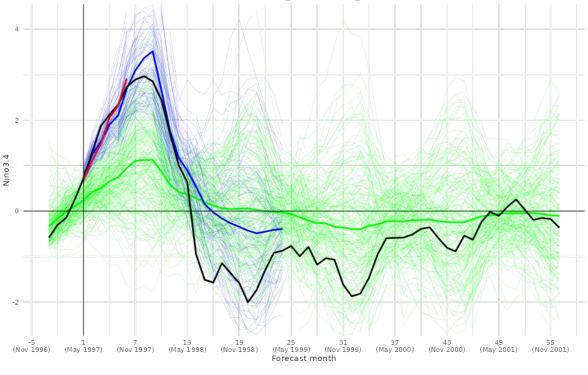






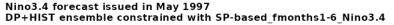


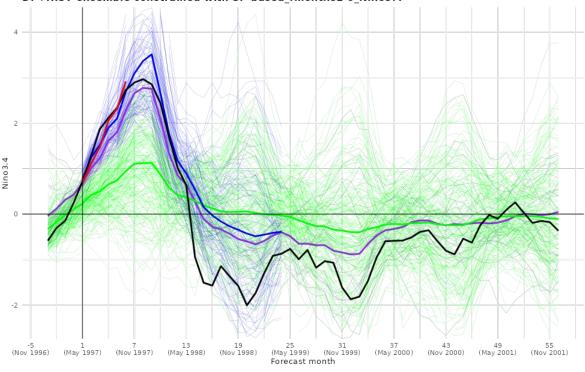




— SP — DP mean — MP mean

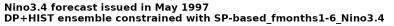


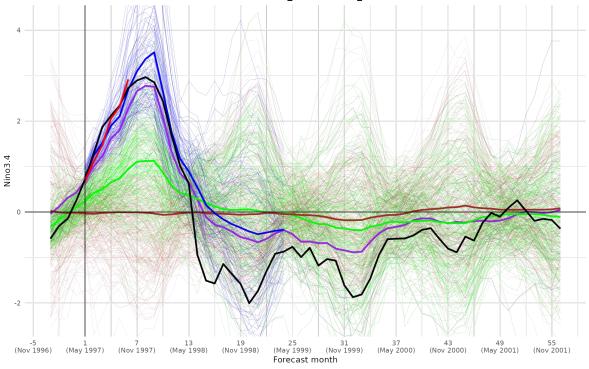




— OBS — SP — DP mean — MP mean — Best mean

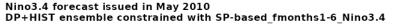






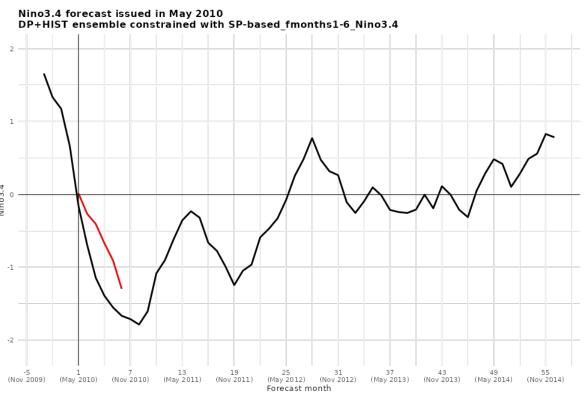
SP - DP mean - MP mean - Best mean - HIST mean





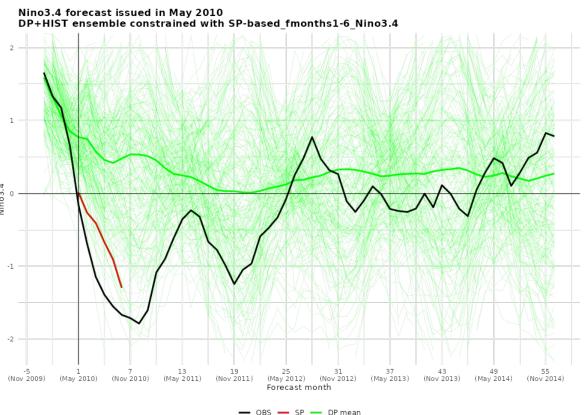




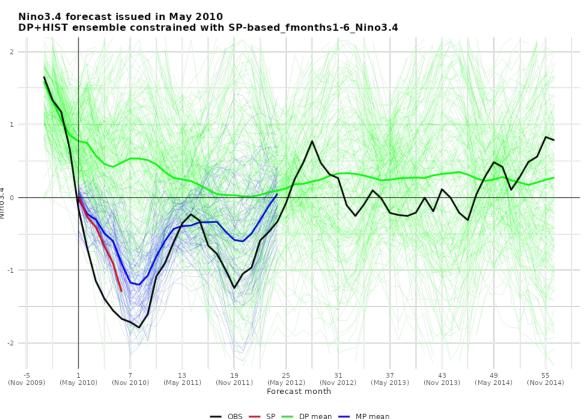


— OBS — SP

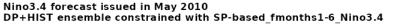


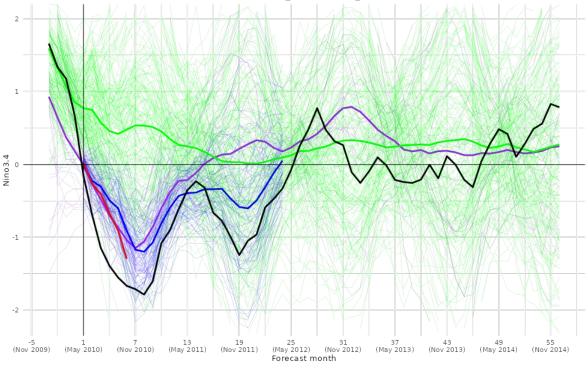






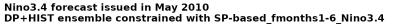


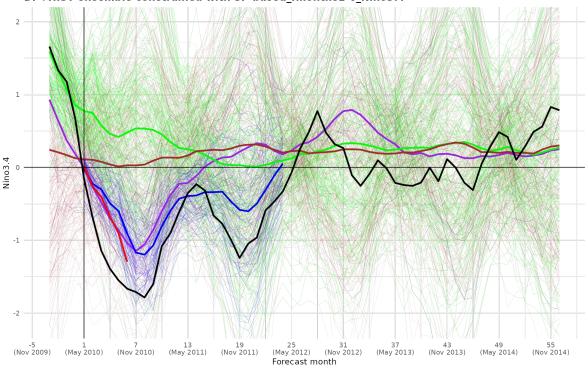




— OBS — SP — DP mean — MP mean — Best mean







SP - DP mean - MP mean - Best mean - HIST mean



Conclusions

- **SP-based constraints** show higher skill than OBS-based constraints
- Initialised MP represents the upper bound of skill achievable by the constraining approaches
- MP are not typically produced in real-time, so constraining approaches offer a cheap alternative (with slightly lower skill) for operational forecasts
- Selecting only DP members provides the highest skill (compared to selecting HIST or DP+HIST members). However, DP availability is lower in real-time, requiring the inclusion of HIST members
 → trade-off between forecast skill and operational feasibility
- During the overlapping period, constrained predictions are as skillful as SP, which suggests that the **constrained predictions can be used since the beginning of the forecast period** (using SP to select the best members)
- Constraining approaches not only improve skill, but also ensures consistency during the transition between timescales
- The **method can be optimised** for specific variables and regions





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UPCLIV Workshop Bologna, November 2025







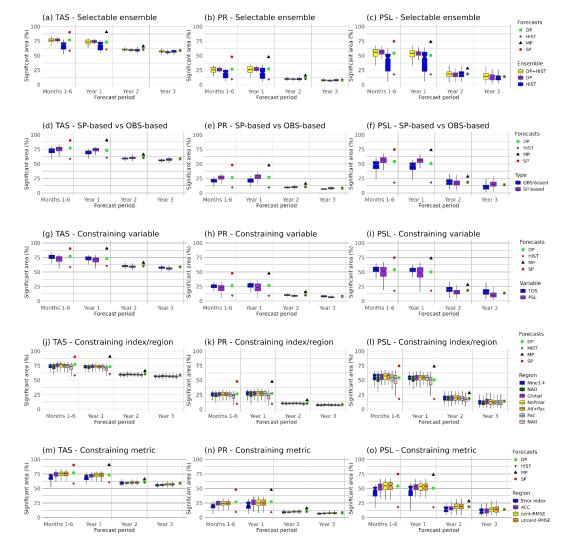
Thank you!

Delgado-Torres, C., Donat, M. G., Pérez-Zanón, N., Torralba, V., Bilbao, R., Bretonnière, P.-A., Samsó-Cabré, M., Soret, A., and Doblas-Reyes, F. J.: Seamless climate information for the next months to multiple years: merging of seasonal and decadal predictions, and their comparison to multi-annual predictions, EGUsphere [preprint], https://doi.org/10.5194/egusphere-2025-3674, 2025.

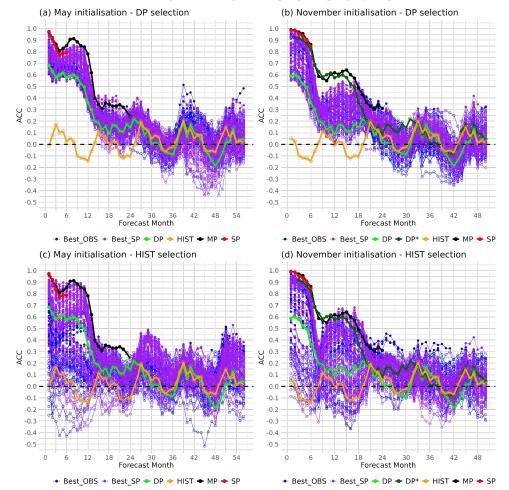
carlos.delgado@bsc.es

Sensitivity analysis to different constraining options





DP vs HIST selection



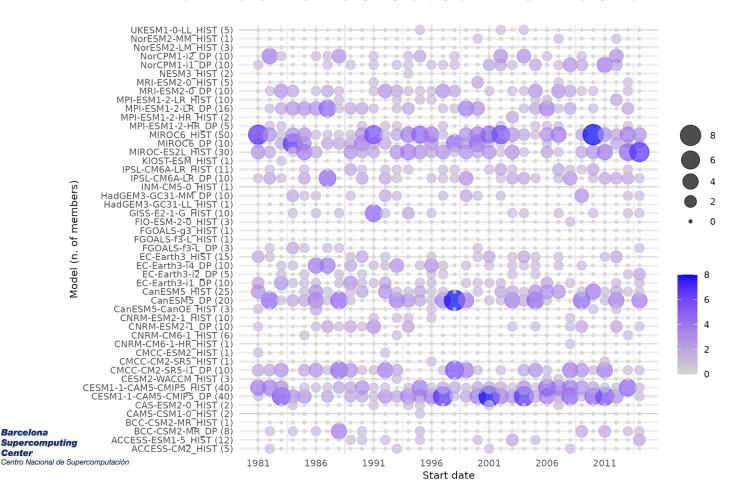
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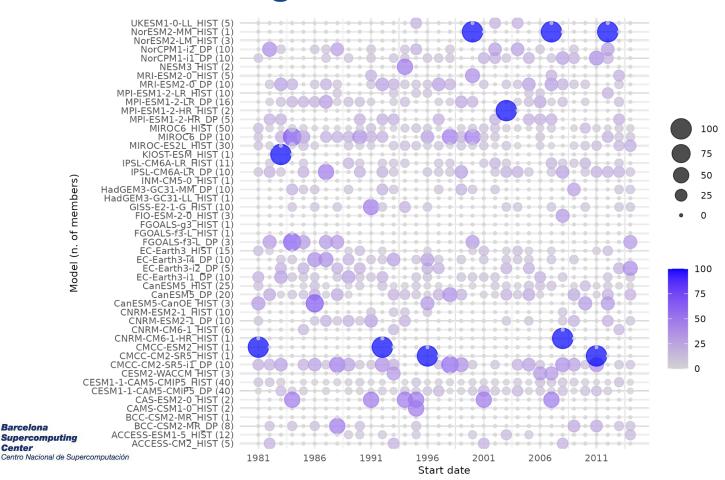
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Number of selected members



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Percentage of selected members



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