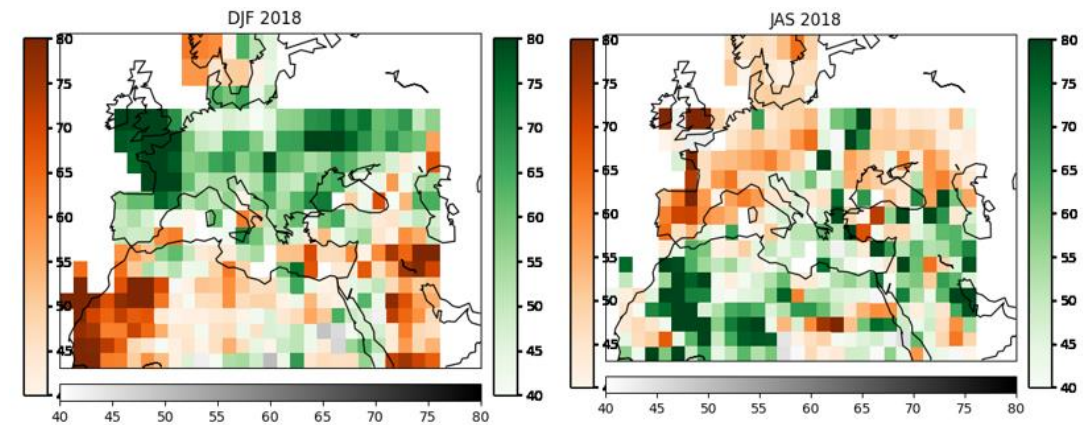
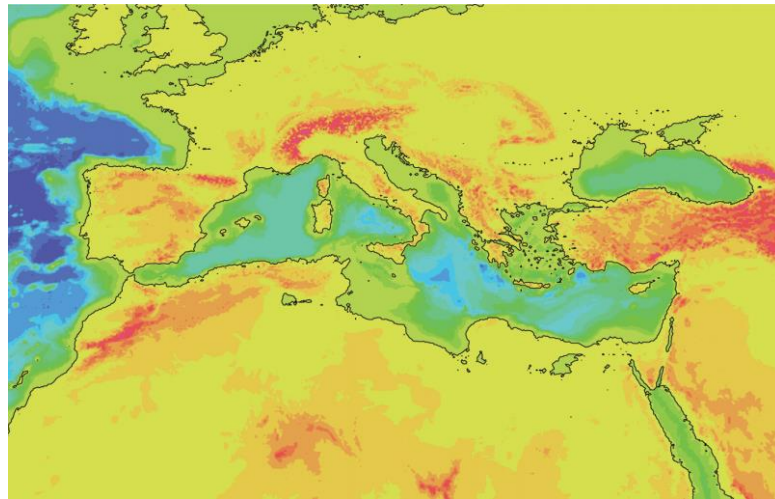


Medscope's Task 2.4 Development of empirical forecast systems

Development of an Empirical Forecast System for the Mediterranean

Rodríguez-Guisado, E; Serrano de la Torre, A.; Sanchez-Garcia, E.; Navascués, B.; Domínguez-Alonso, M.; Rodríguez-Camino, E.

AEMET

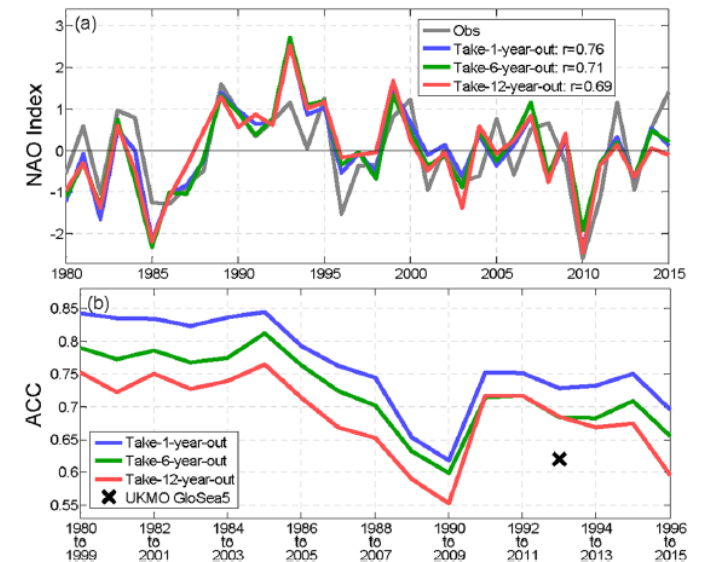


Outline

- Introduction/motivation
- Requirement of MEDSCOPE Task 2.4
- Base line version of Empirical Seasonal Forecasting System
- Verification/ Comparison against Dynamical Systems
- Final remarks
- Future work

Task 2.4 Development of empirical forecast systems

- Dynamical seasonal forecasts still show low skill over Europe/Mediterranean regions
- Computationally-cheap
- Possibility of different ways to combine results with dynamical forecasts
- Good results when predictors are carefully selected (Wang et al, 2017) (tasks 2.1, 2.2, 2.3)
- Development of a preliminar version of the model

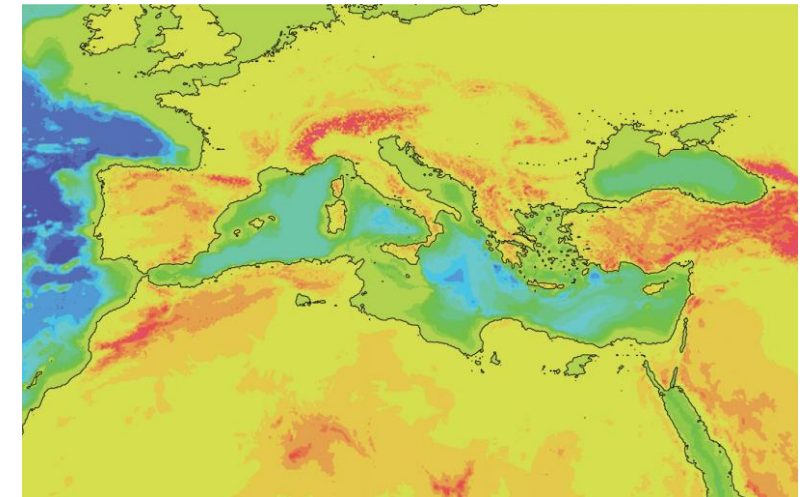
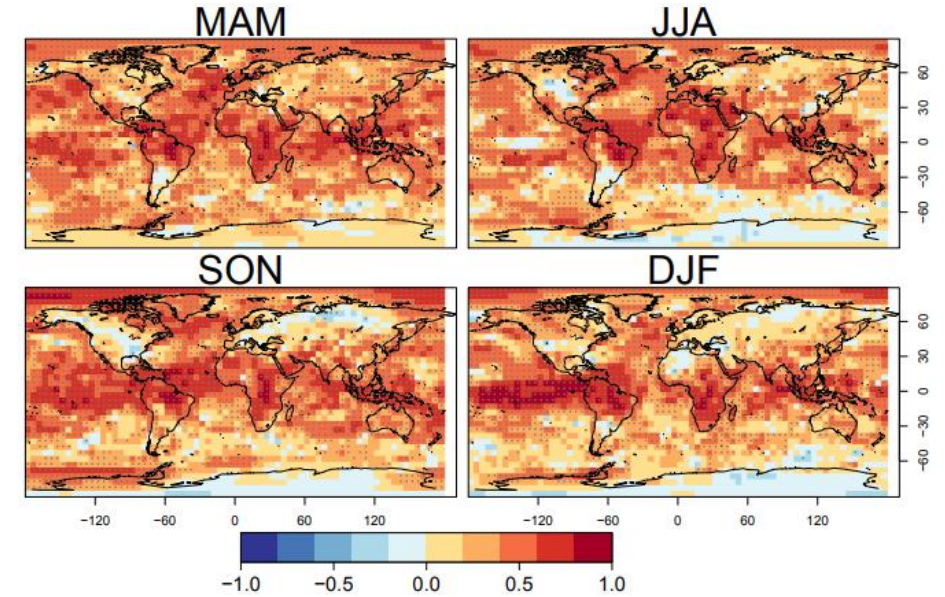


Task 2.4 Development of empirical forecast systems

- First version of empirical model
 - Based on global climate indices
 - Flexible: new predictors should be added easily
 - “Continuous” look: avoid too much noise.

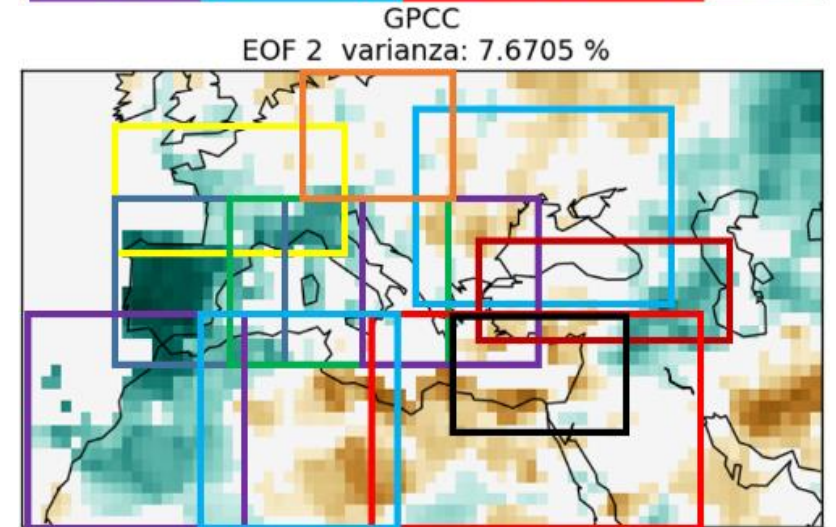
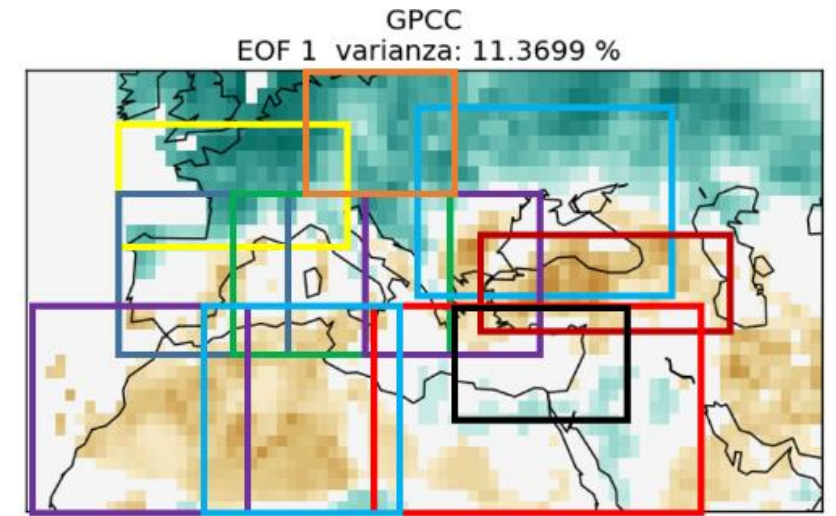


- Multiple linear regression (MLR) selected (similar to Eden et al, 2015), or Ibimet model (Pasqui et al, 2007) Possible problems:
 - Predictors show signal at different areas and times, and can introduce noise at regions where there is no signal from them.
 - Mediterranean area: influenced by many factors, including orographic complexity.
 - Divide in regions to select best predictors for each area
 - MLR results can be noisy (other alternatives are not discarded for final version)



Proposal: divide the area in different regions.

- Selecting areas on a compromise based on patterns of variability and country borders
- Overlap among areas to produce “soft” transitions (weighted interpolations)
- Search of predictors for every area and season
- Partial coincidence in predictors from neighbour areas and consecutive seasons for continuity
- Use of low resolution predictands (GPCC for precipitation and ERA interim for t2m, both 2.5 degrees)



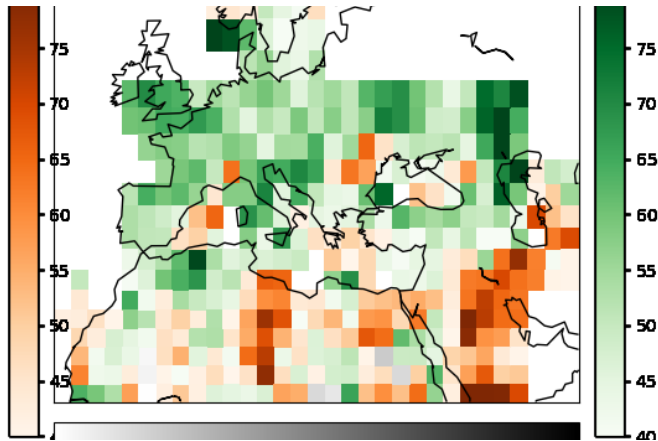
- Testing and selecting predictors:
 - 1979-2017 period
 - Around 30 predictors tested SST-based, global patterns, ice and snowcover based
 - Absolute value and month to month increment tested
 - Up to 12 months lead
 - 2-6 months moving average.

- 4-8 predictors selected for each area and season.
- Around half of them should coincide among neighbour areas and between consecutive seasons
- Avoid collinearity

	GPCC	EFM										FMA										MAM									
		-2	-3	-4	-5	-6	-7	-8	-9	-10	-2	-3	-4	-5	-6	-7	-8	-9	-10	-2	-3	-4	-5	-6	-7	-8	-9	-10			
EA	1M	9	55	18	12	0	9	3	0	18	15	0	15	6	12	21	0	0	3	0	3	0	9	6	3	52	15	6			
	2M	0	45	9	39	9	0	12	3	9	0	15	15	9	0	15	6	12	9	0	0	3	0	0	9	18	6	12			
	3M	0	0	15	6	21	6	0	9	0	0	0	21	18	3	6	9	6	18	0	0	0	3	0	6	15	0	0			
	4M	0	0	0	6	0	30	3	6	6	0	0	18	12	3	9	9	6	0	0	0	0	0	6	0	0	0	0			
	5M	0	0	0	0	3	6	15	0	0	0	0	0	15	6	12	6	6	0	0	0	0	0	6	9	0	0	0			
	6M	0	0	0	0	0	9	9	6	0	0	0	0	0	0	18	12	12	9	0	0	0	0	0	0	6	0	0			
incr_EA	1M	9	48	6	15	12	6	3	6	3	12	6	15	18	15	15	0	3	9	3	3	3	12	12	30	42	12	3			
	2M	0	24	58	6	15	3	0	9	0	0	6	6	12	6	6	0	12	0	0	0	0	6	30	12	24	0				
	3M	0	0	45	36	6	21	3	15	6	0	21	12	27	0	6	18	24	0	0	0	9	6	9	36	18	0	21			
	4M	0	0	0	6	30	21	15	3	15	0	0	18	0	6	9	12	3	0	0	0	0	0	15	0	3	0				
	5M	0	0	0	0	9	18	9	12	0	0	0	0	27	12	3	6	6	0	0	0	0	0	0	33	12	9	6			
	6M	0	0	0	0	0	6	18	3	12	0	0	0	0	9	6	3	0	0	0	0	0	0	0	12	12	6	3			
EA/WR	1M	6	6	0	6	0	6	9	39	3	3	6	3	6	15	6	9	6	3	0	0	3	3	45	21	3	0	6			
	2M	0	6	3	9	0	3	0	21	15	0	9	6	0	6	15	12	18	9	0	0	3	0	18	24	0	0	3			
	3M	0	0	6	3	3	0	0	15	21	0	0	6	0	0	9	6	15	12	0	0	3	0	12	6	9	6	0			
	4M	0	0	0	6	3	3	0	0	15	0	0	0	0	3	3	6	3	0	0	0	0	0	9	9	6	12	12			
	5M	0	0	0	0	6	3	0	0	0	0	0	0	0	3	6	0	0	3	0	0	0	0	0	6	6	3	9			
	6M	0	0	0	0	0	6	0	0	0	0	0	0	0	0	3	0	0	0	0	0	0	0	0	0	6	6	3			
incr_EA/WR	1M	3	6	0	0	3	0	3	6	15	0	15	6	9	3	9	3	6	6	3	0	9	24	36	21	3	0	0			
	2M	0	3	0	0	3	9	21	9	0	0	9	3	3	12	15	12	21	9	0	6	6	9	21	12	9	9	6			
	3M	0	0	12	3	0	18	12	0	15	0	0	3	3	0	18	12	21	0	0	0	0	9	6	6	9	6	0			
	4M	0	0	0	6	3	6	27	6	36	0	0	0	9	9	3	6	24	3	0	0	0	27	21	0	9	12	24			
	5M	0	0	0	0	18	3	27	3	3	0	0	0	0	6	9	9	0	6	0	0	0	0	18	0	0	0	0			
	6M	0	0	0	0	0	9	3	27	0	0	0	0	0	0	9	12	15	15	0	0	0	0	0	6	0	0	6			
SCAND	1M	3	0	12	0	6	18	52	0	33	15	0	3	27	9	0	3	15	18	3	3	18	0	3	15	0	3	3			
	2M	0	0	3	0	0	9	30	9	12	0	12	3	12	24	0	0	9	6	0	6	12	3	6	9	3	0	0			
	3M	0	0	0	0	15	9	0	15	0	0	0	9	6	15	27	0	0	6	0	0	15	9	6	6	6	0	0			
	4M	0	0	0	0	12	15	0	3	0	0	0	0	0	9	15	6	3	6	0	0	0	18	6	9	6	0	0			
	5M	0	0	0	0	0	9	0	3	0	0	0	0	0	0	9	6	3	6	0	0	0	0	18	9	9	0	0			
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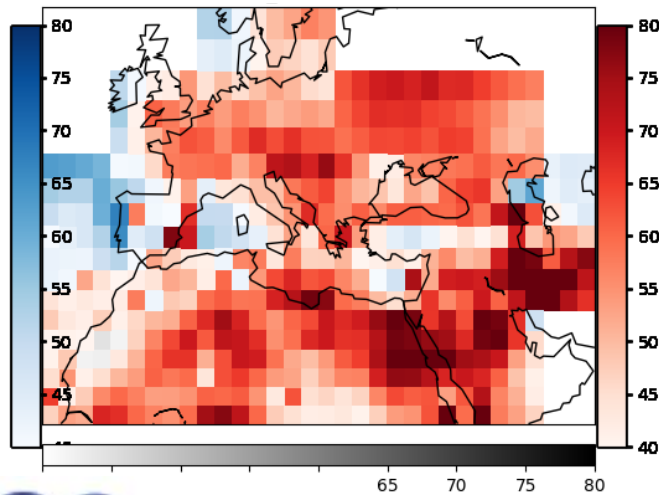
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EFM	LAAO_S EA_ON TNA_JJIR L_SO_Ns_SQ DMI_SON	LAAO_JJAS EA_JJAS EA_ON DMI_SON SOI_JJ	EA_SON LAAO_JJAS EA_JJAS SOI_JJ L_SO_EuA_JJIR	LAAO_SQ SOI_JJ LAAO_S EA_ON L_SO_EuA_Q DMI_LAO_S L_SOI_O	EA_ON L_SO_Ns_Q SOI_JJ EA_ON L_SO_EuA_Q DMI_LAO_S L_SOI_O	NAC-N L_SO_Ns_Q SOI_JJ SOI_JJ L_NIRs4_N LAAO_SON	NAC_JA L_SO_Ns_SQ TNA_ON L_NIRs4_N EAWR_O
FMA	LAAO_S NAC-OND L_NIRs4_N EA_OND TNA_D TNA_JJIR TNA_D L_SO_Ns_SQ DMI_SON	LAAO_JJAS NAC-OND EA_OND TNA_D TNA_D DMI_SON SOI_JJ L_NIRs4_N	LAAO_JJAS NAC-OND L_NIRs4_N TNA_D TNA_D SOI_JJ L_SO_EuA_JJIR	SOI_JJ LAAO_S L_SO_Ns_Q L_SOI_O	L_NIRs4_N L_SO_Ns_Q SOI_JJ EA_ON DMI_LAO_SON LAAO_SON	NAC-ND L_SO_Ns_SQ SOI_JJ SOI_JJ L_NIRs4_ND TNA_ND	NAC_JA L_SO_Ns_SQ TNA_ON L_SOI_O L_NIRs4_N EAWR_O
MAM	NAC-ONDE L_NIRs4_N TNA_JJIR TNA_DE L_SO_Ns_SQ DMI_SON	NAC-ONDE L_NIRs4_N TNA_JJIR TNA_DE L_TNA_ND DMI_SON SOI_JJ	NAC-ONDE L_NIRs4_N TNA_JJIR TNA_DE SOI_JJ	SOI_JJ LAAO_SON L_SO_Ns_Q L_SOI_O	L_NIRs4_N TNA_JJIR TNA_DE TNA_DE EA_ON DMI_LAO_SON LAAO_SON	NAC-ND L_SO_Ns_SQ SOI_JJ SOI_JJ L_NIRs4_ND TNA_ND	NAC_JA L_SO_Ns_SQ TNA_ON L_SOI_O DMI EAWR_O
AMJ	NAC-ONDE L_NIRs4_N TNA_DE DMI_SON L_NAO_EF	NAC-ONDE TNA_JJIR TNA_DE DMI_SON L_TNA_ND L_NAO_EF SOI_JJ	NAC-ONDE L_NIRs4_N TNA_DE TNA_DE L_NAO_EF SOI_JJ	SOI_JJ TNA_DE TNA_AS NAC-ONDE LAAO_NDE	L_NIRs4_N TNA_JJIR TNA_DE EA_ON	NAC-ND L_SO_Ns_SQ L_NAO_EF L_SOI_O L_NIRs4_ND TNA_ND	L_NIRs4_NDE L_SOI_O DMI EAWR_O
MJJ	TNA_JJIR TNA_DE L_NAO_EF TSA_FM	TNA_JJIR NAC-ONDE L_EA/WR_FM TNA_DE L_TNA_NDE L_NAO_EF L_EA_O	L_SO_EuA_JJIR TSA_FM L_NAO_EF L_SO_EuA_EF	TNA_E TNA_AS NAC-ONDE LAAO_NDE SAM_O	TNA_JJIR EA_ON LAAO_NDE L_NAO_EF SAM_O	L_SO_Ns_SQ L_NAO_EF L_SOI_O TNA_DEF TSA_FM	L_NIRs4_NDE L_SOI_O DMI EAWR_O PDO_N

- Example results:
 - Precipitation



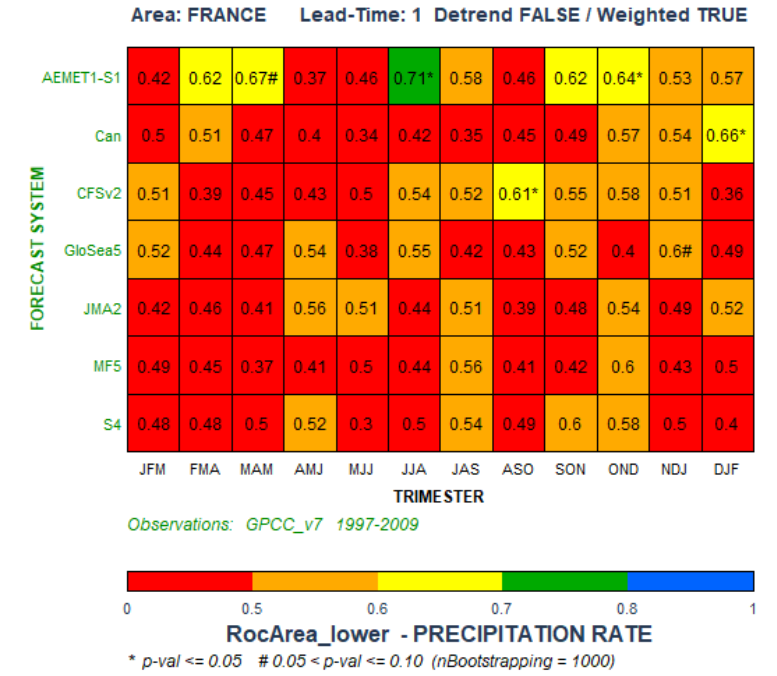
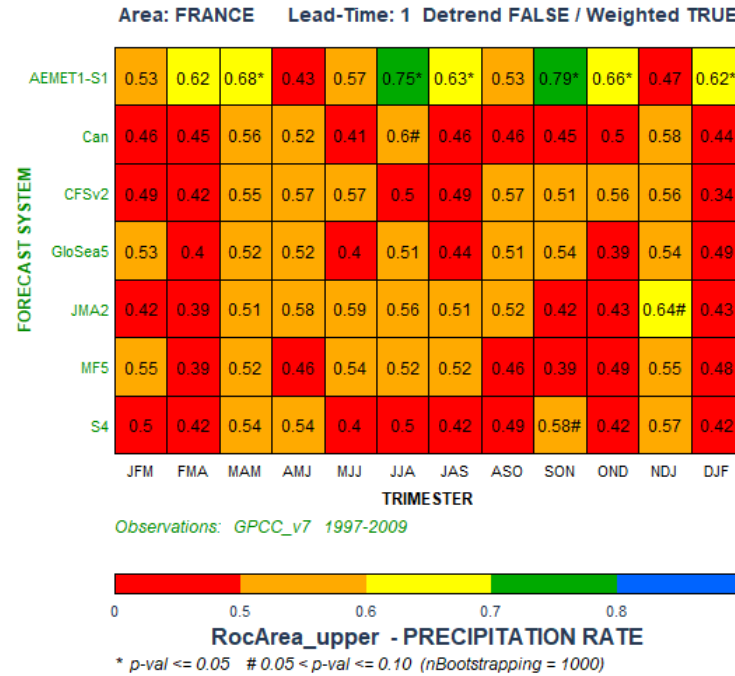
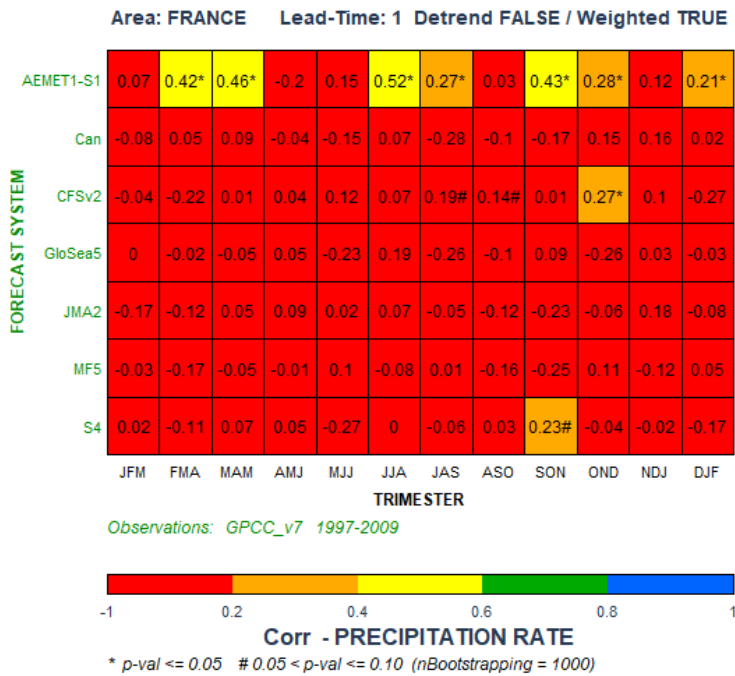
- Synoptic structures observed.
- Not too much noise
- Borders can still be seen some times
- Currently predictors table constructed manually.
- Currently selection of predictors made for precipitation. Use of the same predictors for temperature

- Temperature

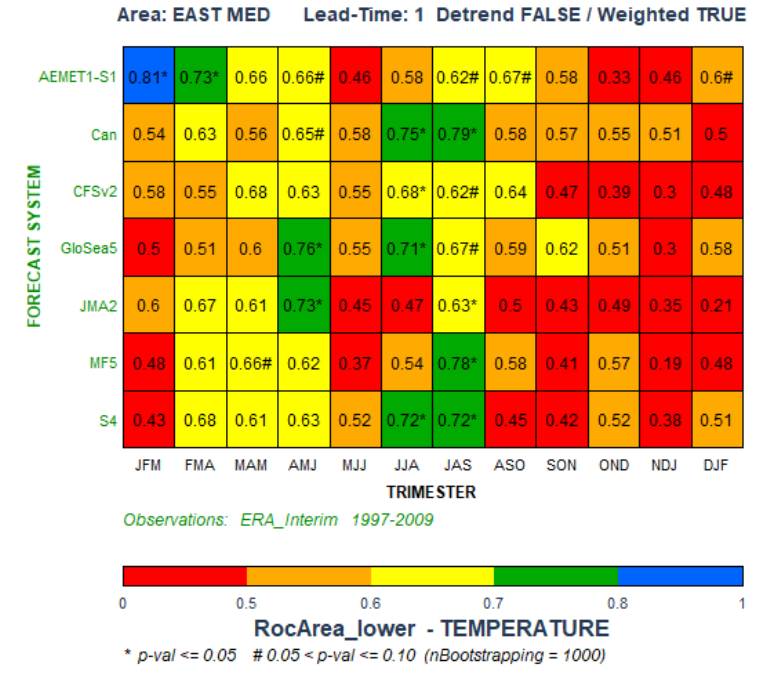
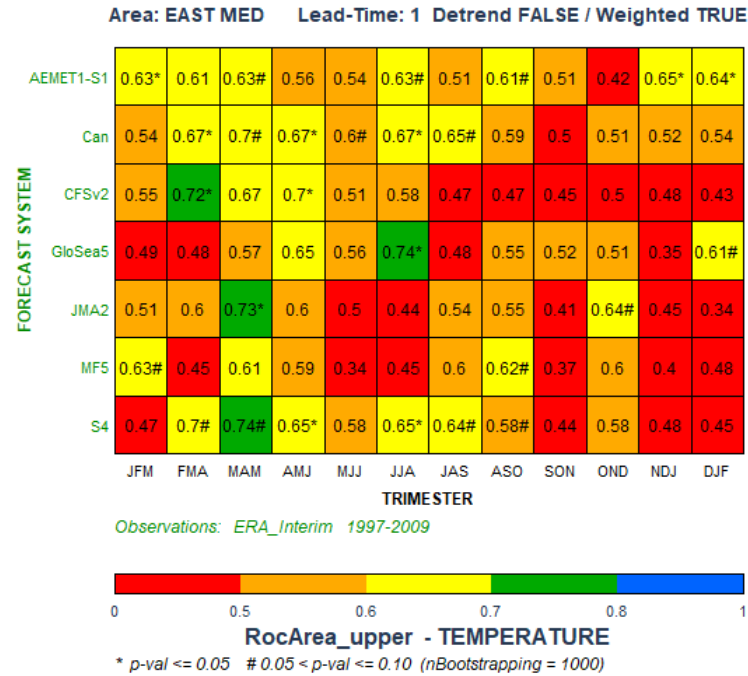
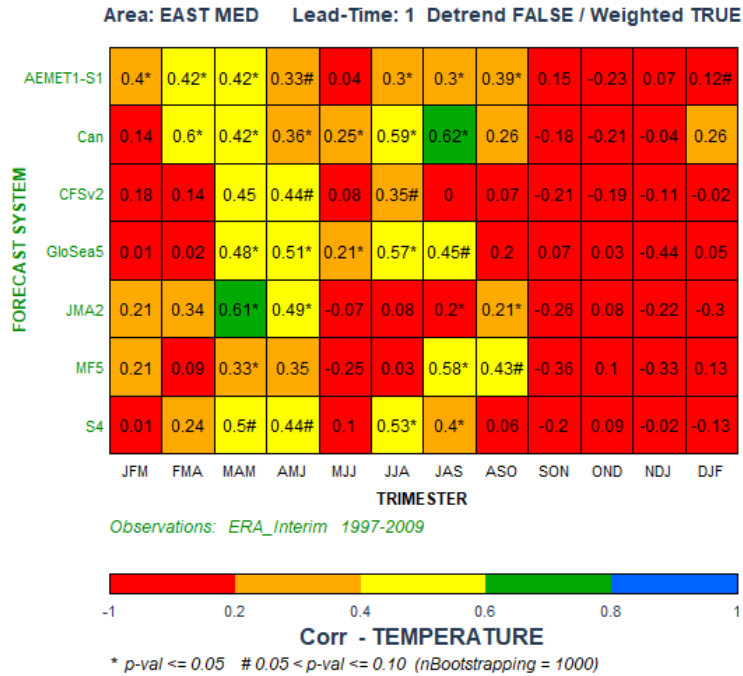


Task 2.4 Development of empirical forecast systems

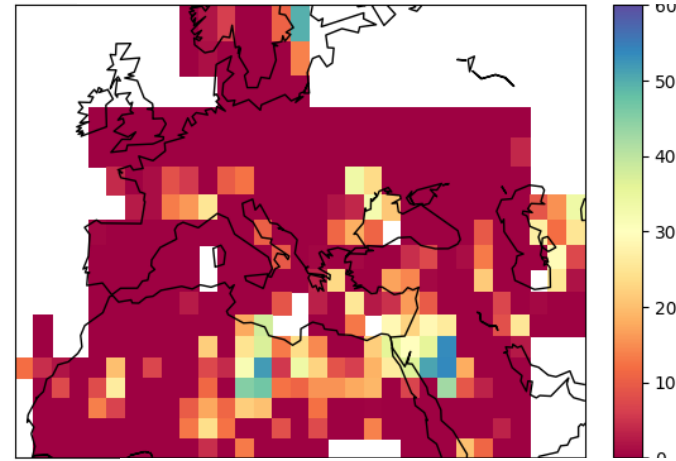
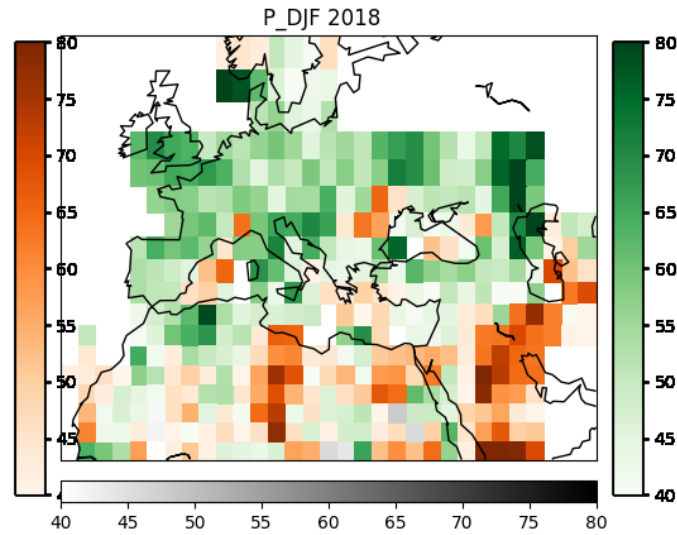
- Verification of empirical system (v1) compared with main seasonal forecast dynamical models :
- Correlation and ROC area for France
Precipitation



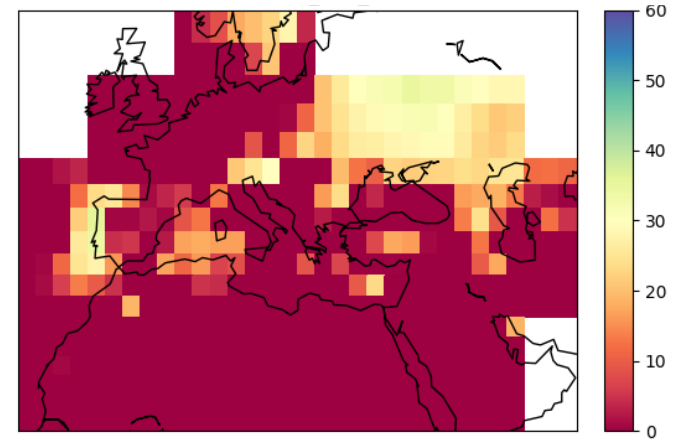
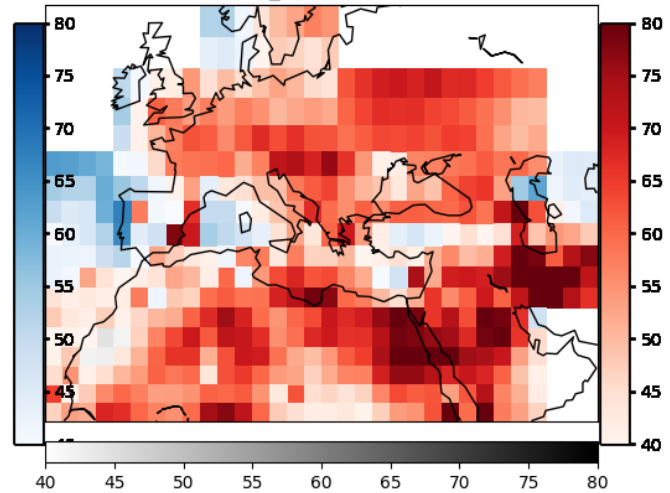
- Verification of zones against main seasonal forecast dynamical models :
- Correlation and ROC area for Eastern Mediterranean: Temperature



- Example results:
- Precipitation



- Temperature



Conclusions:

- Preliminary version of empirical model showed ability to produce synoptic anomaly patterns
- Good results (considering that this is only the first version) for some regions show potential of this type of model.
- Still work to do on selection of predictors for every area.

Next Steps:

- Automatic procedure for selection of predictors
- Exploration of specific predictors for temperature
- Incorporation of snow cover (and other) indices with no available data during the hole period (SAI/SNV not used)
- Incorporation of results from other tasks

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