



On the use of weather generators for the estimation of low-frequency floods under climate change scenarios

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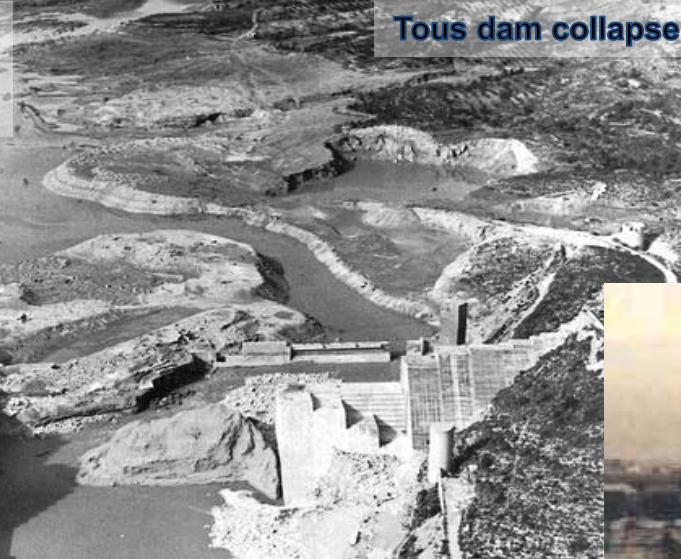
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- Long history of devastating floods



Oct-1982
>500mm in 72 hrs
8 fatalities



Tous dam collapse



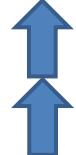
Carcaixent



Valencia - 1957

- But also many recent ones



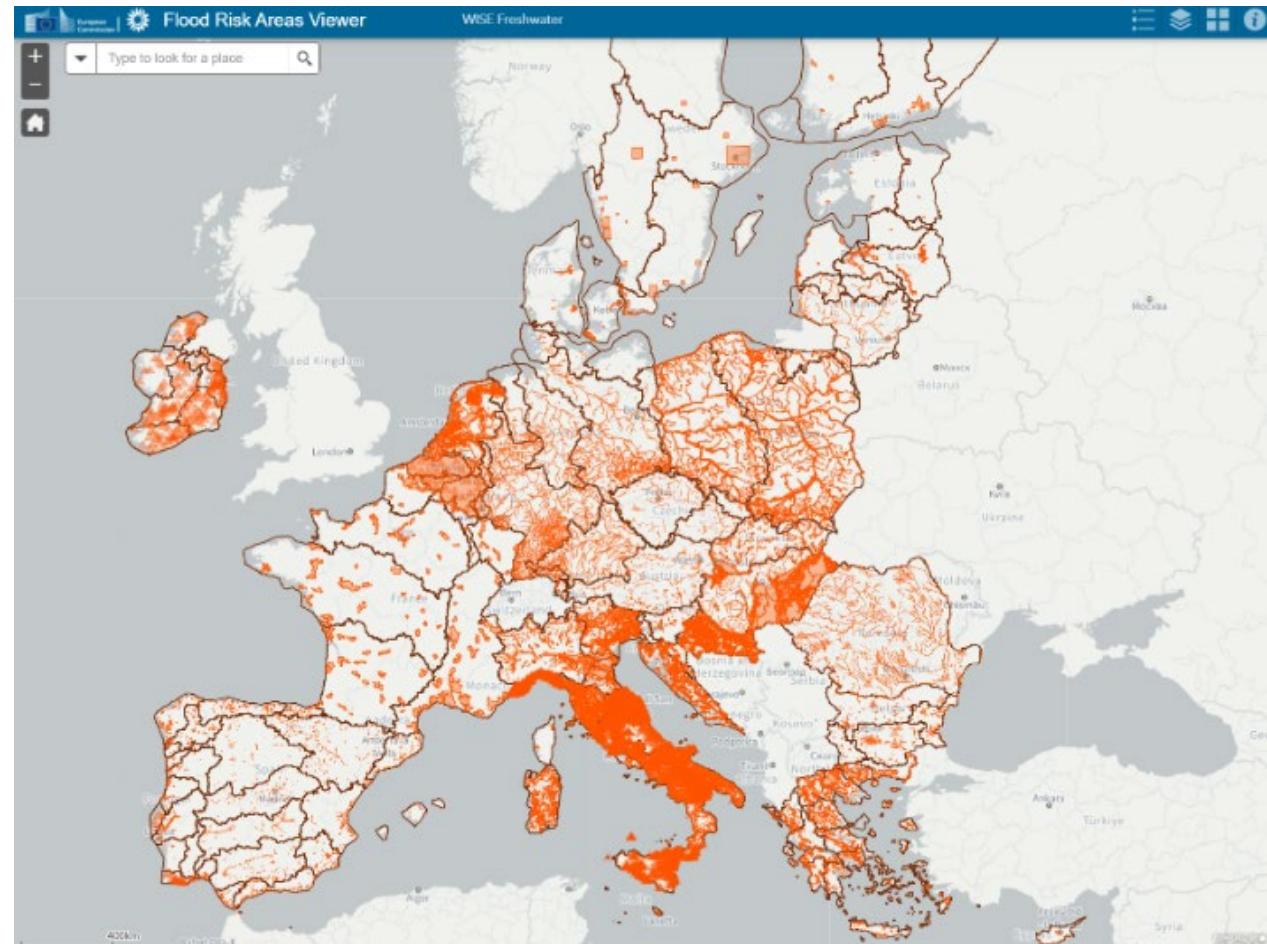
Climate change:  Frequency
Magnitude



Change in heavy rain (%)

Heavy rain is defined as 95th percentile intensity of total rain events

<-25 -25 to -15 -15 to -5 -5 to 5 5 to 15 15 to 25 25 to 35 >35



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On the use of weather generators for the estimation of low-frequency
floods under climate change scenarios

- Importance of extreme floods studies
 - Hydraulic infrastructures design (Spanish Dam Safety Standards, 2023)
 - Flood maps elaboration (Directive 2007/60 CE)
 - Climate change

Develop a new methodology based on stochastic weather generators (weather generator + fully-distributed hydrological model + integration of information) for an adequate estimation of extreme floods (low-frequency) under climate change scenarios

- Traditionally:
 - Q_{obs} fit to a cdf
 - Design Storm
- Synthetic generation of precipitation (+ hydrological models):

➤ Stochastic processes

❖ Single storms

❖ Storm transposition

Short length of hydrometeorological records

Initial state

❖ Clustered point process N-S

Continuous Synthetic Simulation

➤ Stochastic simulation + Hydrological Model

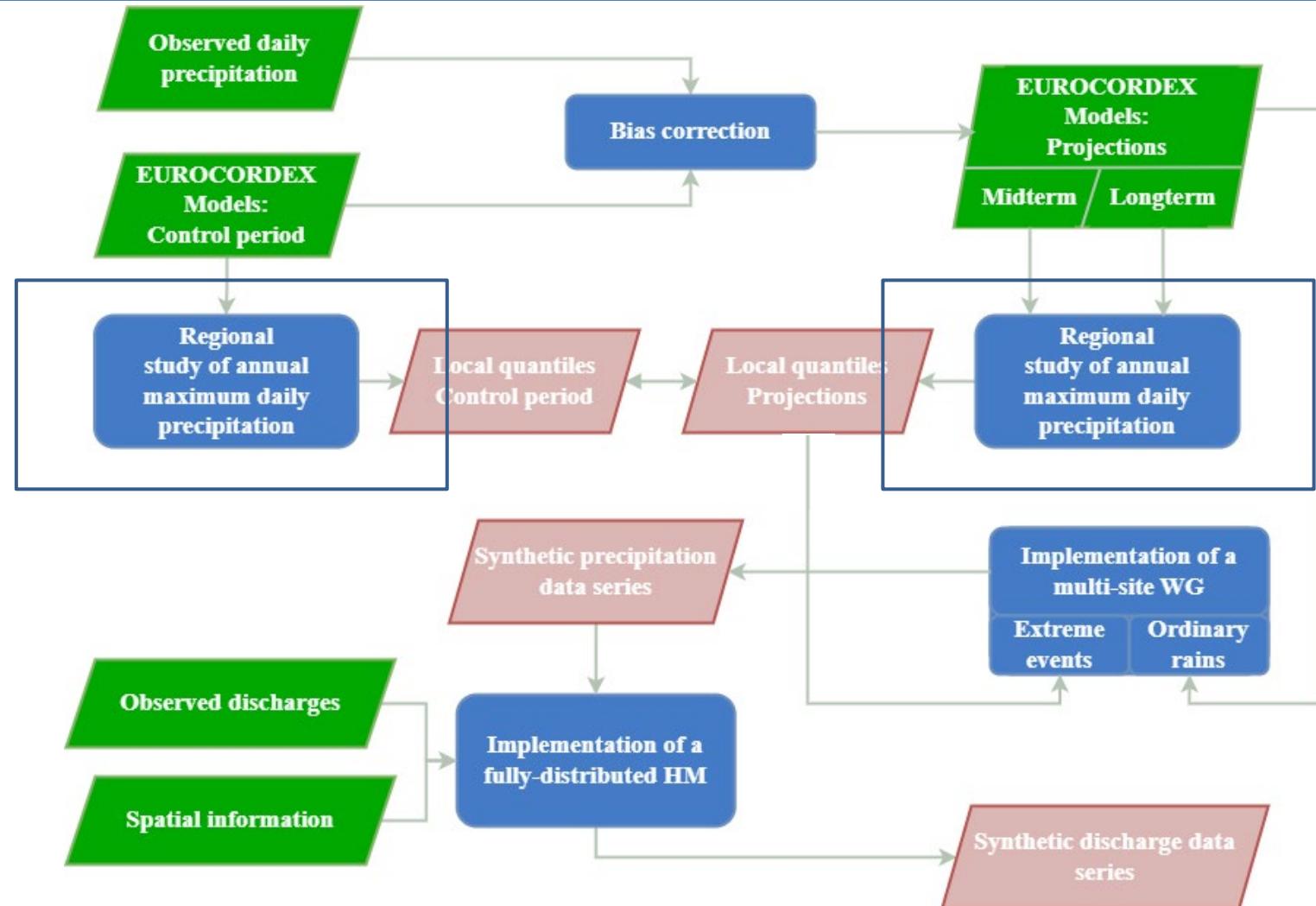
Drawbacks:

➤ Reproduction of extremes in normal conditions of available information

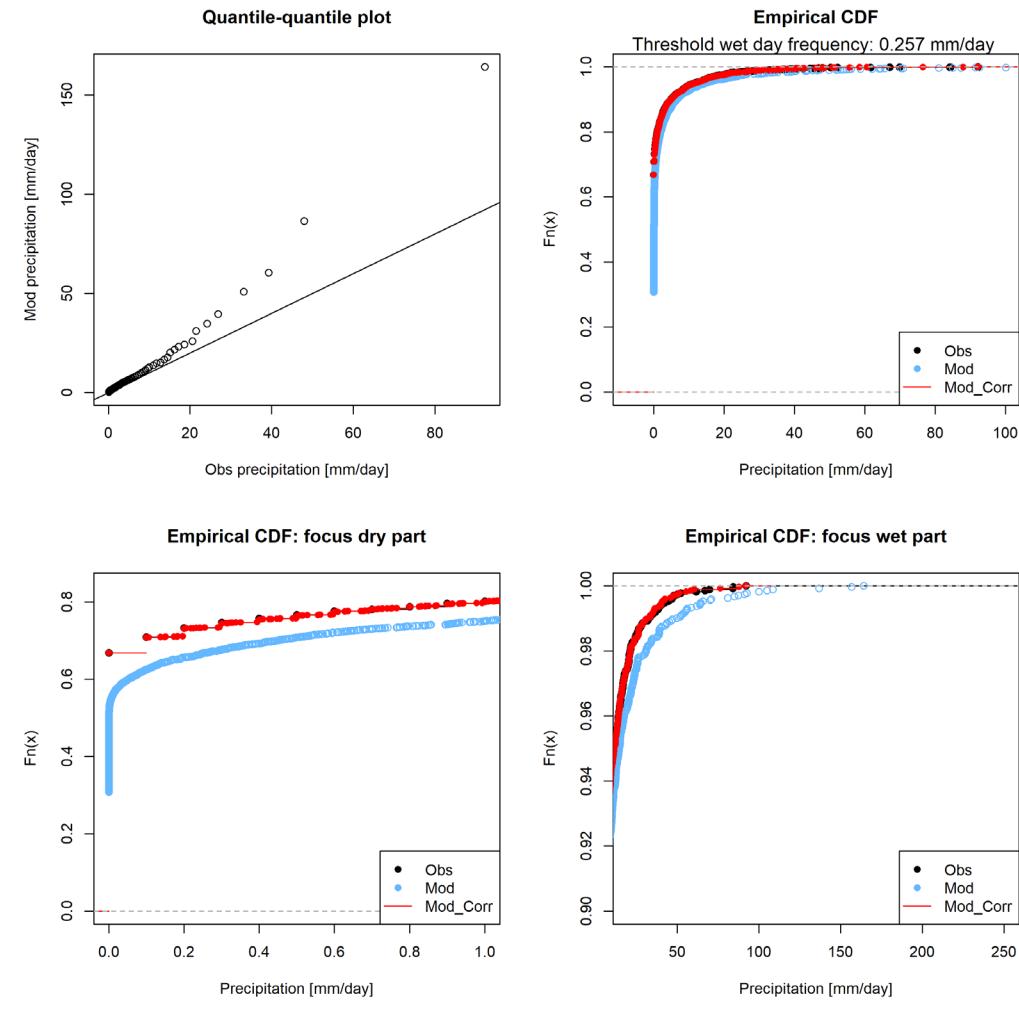
➤ Subdaily weather generators still complex and with high computational requirements

- ✓ Short length of observations
- ✓ Spatio-temporal distribution of the storm
- ✓ Initial conditions of the basin

Beneyto et al, (2024)



- Try P from the 12 climate models with the data from Spain02-v5.
- Based on the non-parametrical statistical transformation of the empirical quantiles (*Gudmundsson et al., 2012*).
- Implemented independently for each season and each grid: December - February, March - May, June - August, and September - November
- Dry and wet day frequency also corrected following the approach proposed by *Themeßl et al. (2012)*.



- L-moments estimation
- Homogeneity analysis (discordance)
- Selection of the regional cdf
- Obtention of the local quantiles

$$X_{i,T} = X_{R,T} \cdot \bar{X}_i$$

where $X_{i,T}$ is the quantile of return period T at location i ,

$X_{R,T}$ is the regional quantile of return period $|T|$

\bar{X}_i is the mean of the registered data at location i .

(Hosking & Wallis, 1993, 1997) (Dalrymple, 1960)

$$PWM_{p,r,s} = E\{X^p \cdot F_X(x)^r \cdot [1 - F_X(x)]^s\}$$

Greenwood et al. (1979)

$$\beta_r = \frac{1}{n} \cdot \sum_{j=r+1}^n \frac{(j-1) \cdot (j-2) \cdot \dots \cdot (j-r)}{(n-1) \cdot (n-2) \cdot \dots \cdot (n-r)} \cdot x_j \quad (\text{Hosking y Wallis, 1997})$$

$$\lambda_1 = \beta_0$$

$$\lambda_2 = 2 \cdot \beta_1 - \beta_0$$

$$\lambda_3 = 6 \cdot \beta_2 - 6 \cdot \beta_1 + \beta_0$$

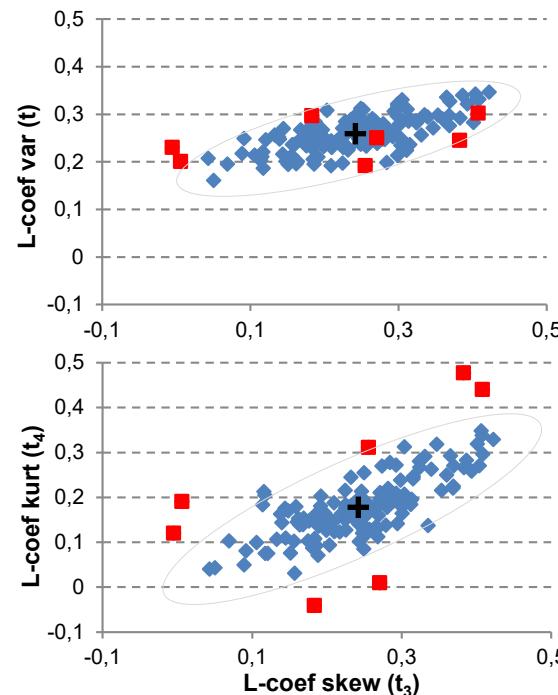
$$\lambda_4 = 20 \cdot \beta_3 - 30 \cdot \beta_2 + 12 \cdot \beta_1 - \beta_0$$

$$t = \frac{\lambda_2}{\lambda_1} \quad t_3 = \frac{\lambda_3}{\lambda_2} \quad t_4 = \frac{\lambda_4}{\lambda_2}$$

t : L-Coefficient of Variation (L-CV)

t_3 : L-Coefficient of Skewness (L-CS)

t_4 : L-Coefficient of Kurtosis (L-CC)



Weather Generator GWEX (Evin et al., 2018)

- Extended Generalized Pareto Distribution (E-GDP)

-> heavy-tailed

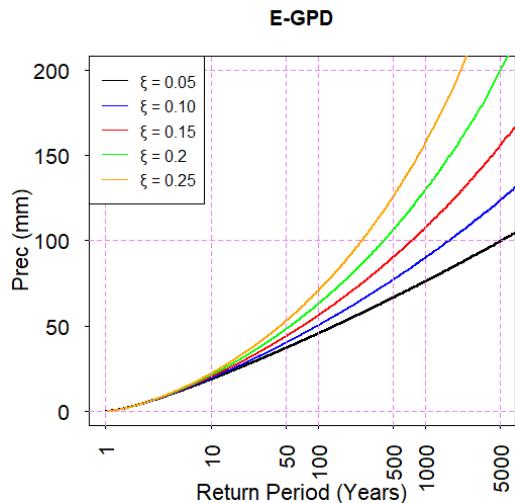
$$F(x; \lambda) = \left[1 - \left(1 + \frac{\xi x}{\sigma} \right)_+^{-1/\xi} \right]^k, x > 0$$

- Parameter estimation

$\sigma, \kappa,$ } From observations

$\xi,$ From more robust studies

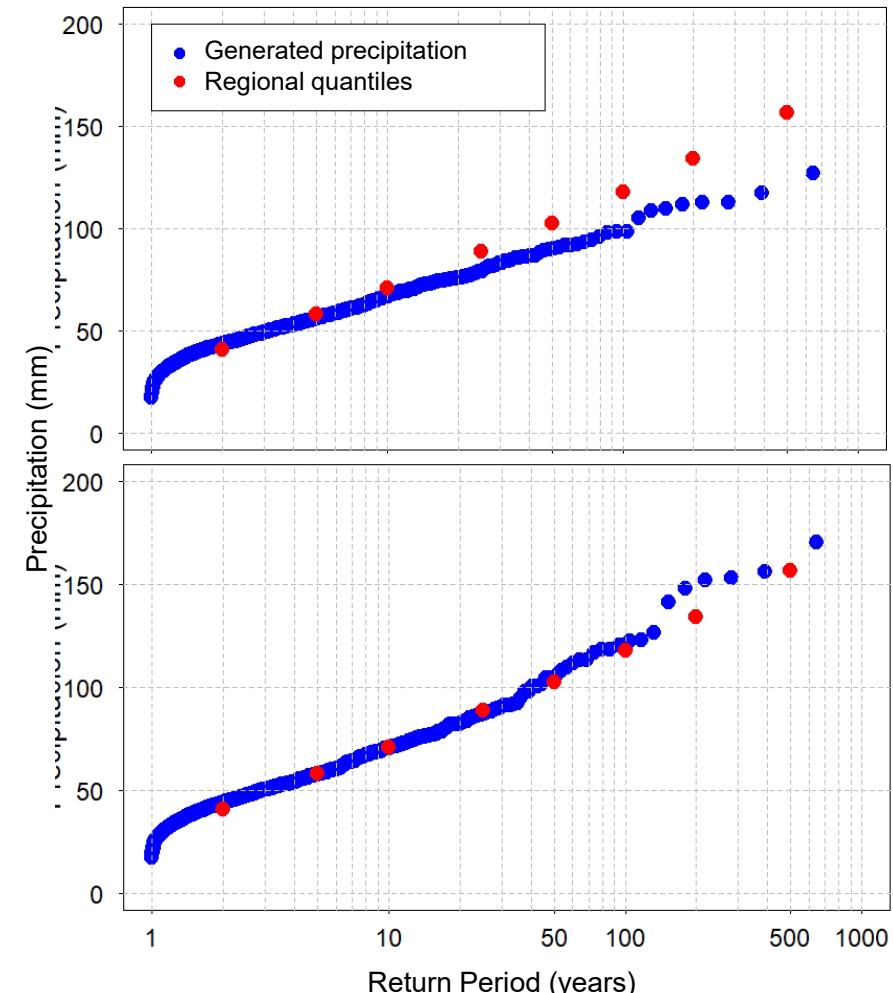
- 3-day aggregation



Beneyto et al, (2020)

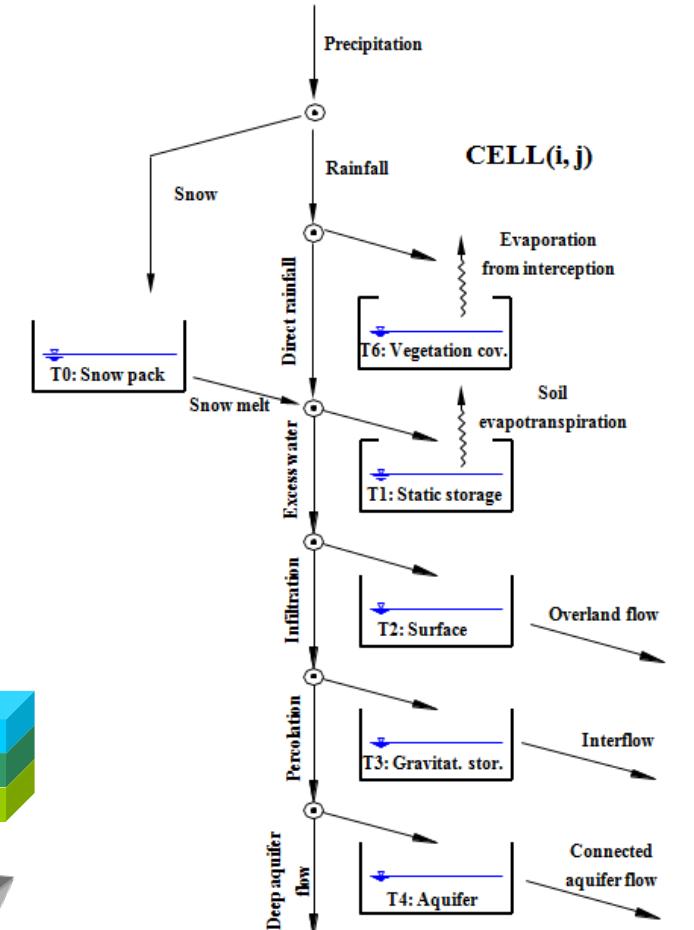
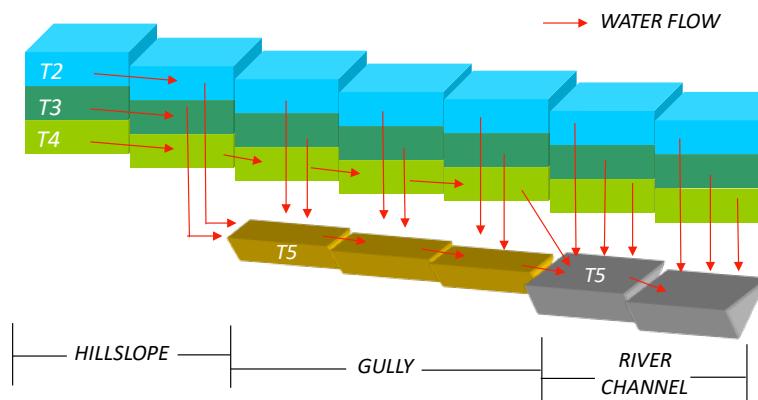
Beneyto et al, (2023a)

Beneyto et al, (2023b)



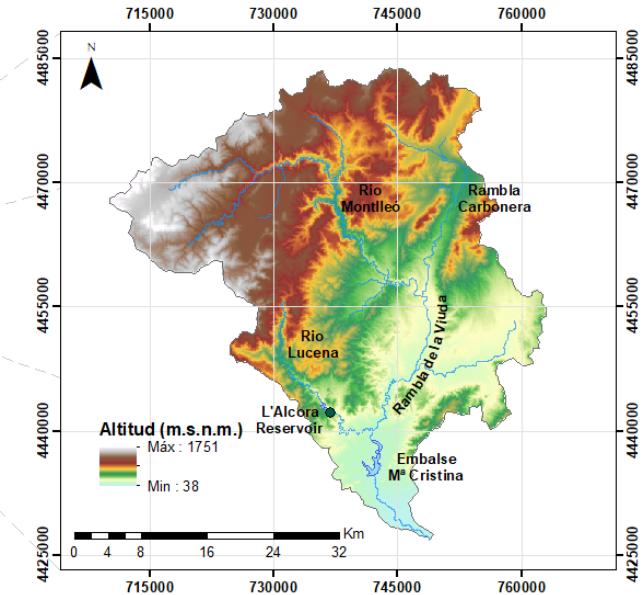
□ **Distributed in space:**

- Constantly developed by our group since 1994
- Integral model
- Parsimonious
- Reproduces the spatial variability of hydrological cycle
- Uses all spatial information available
- Gives results at any point



Esquema conceptual del modelo TETIS
a escala de celda

- **Rambla de la Viuda**: ephemeral river
- Approx. area: 1,500 km²
- Mediterranean semiarid climate
- High precipitation variability
- Two reservoirs (M^a Cristina y Alcora)



□ Hydrometeorological information

- Precipitation
 - Grid Spain02-v5: 20 pluviometers + thermometers: 1951-2015 (66 yrs)
- Flow Gauges
 - SAIH Júcar

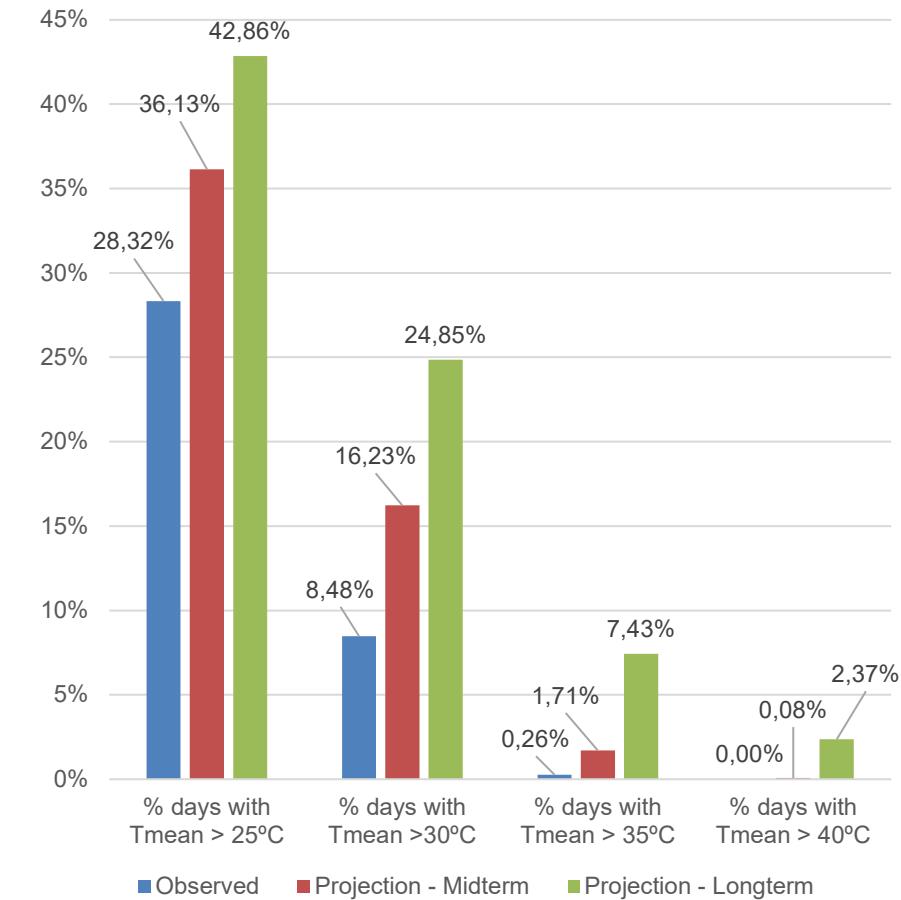
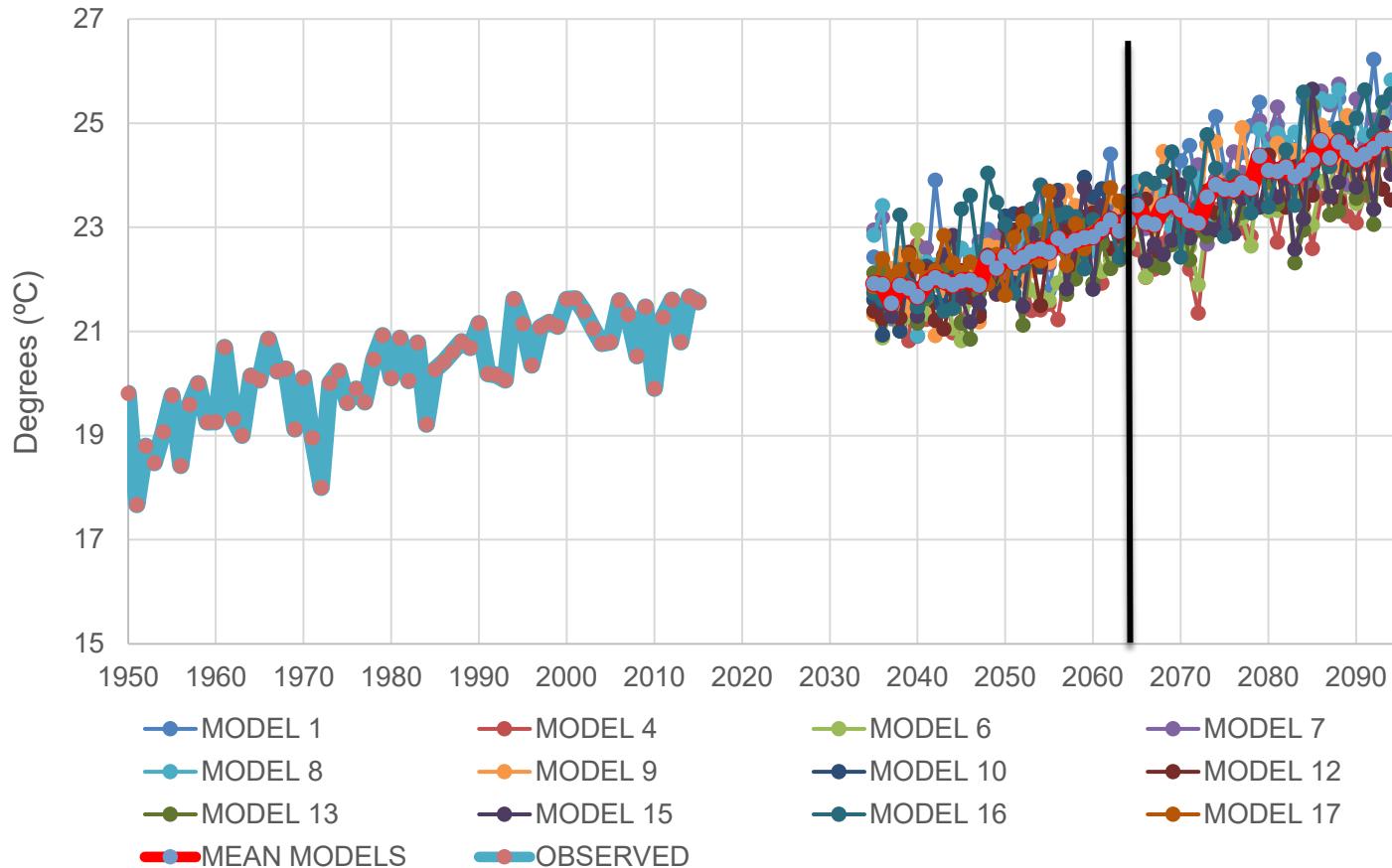
	Data series length (full years)	Period	
		Start	End
M ^a Cristina (Reservoir)	59	1/10/1959	17/12/2018
Alcora (Reservoir)	56	1/10/1959	30/09/2015
Vall d'Alba	15	13/05/2004	17/12/2018
Monleon	14	1/11/2005	20/12/2018

- Climate projections
- 12 Models EUROCORDEX Project
 - Control Period: (1971-2000)
 - Midterm Projection: (2035-2064)
 - Longterm Projection: (2065-2094)

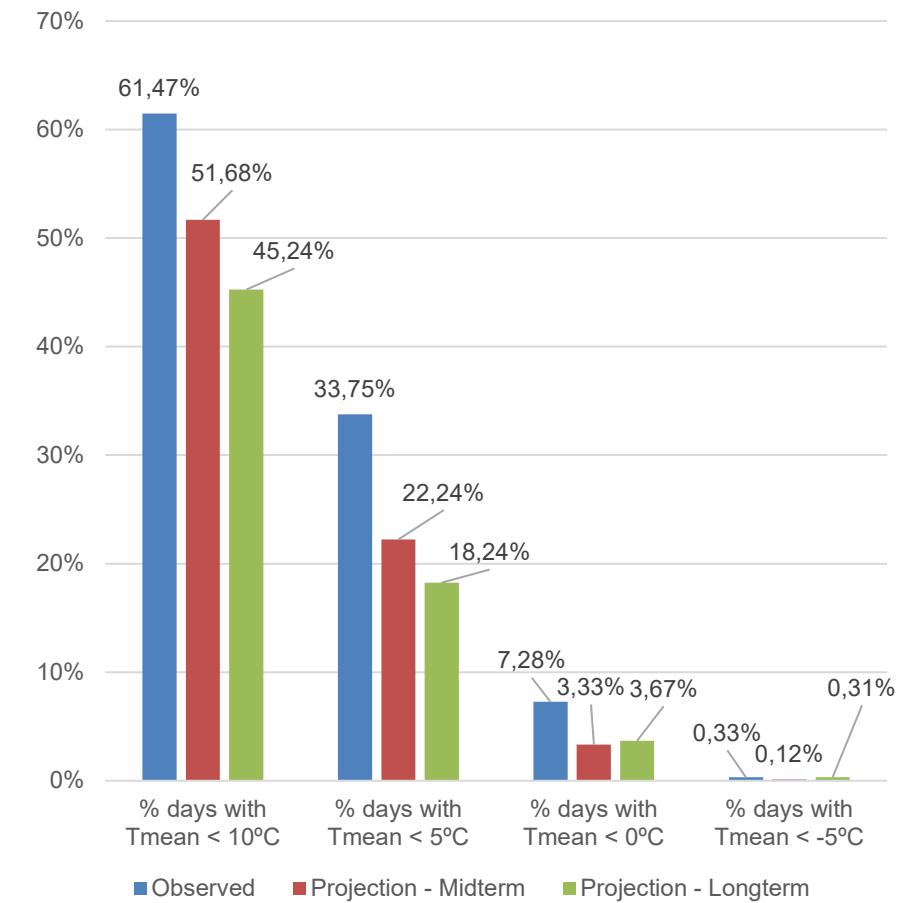
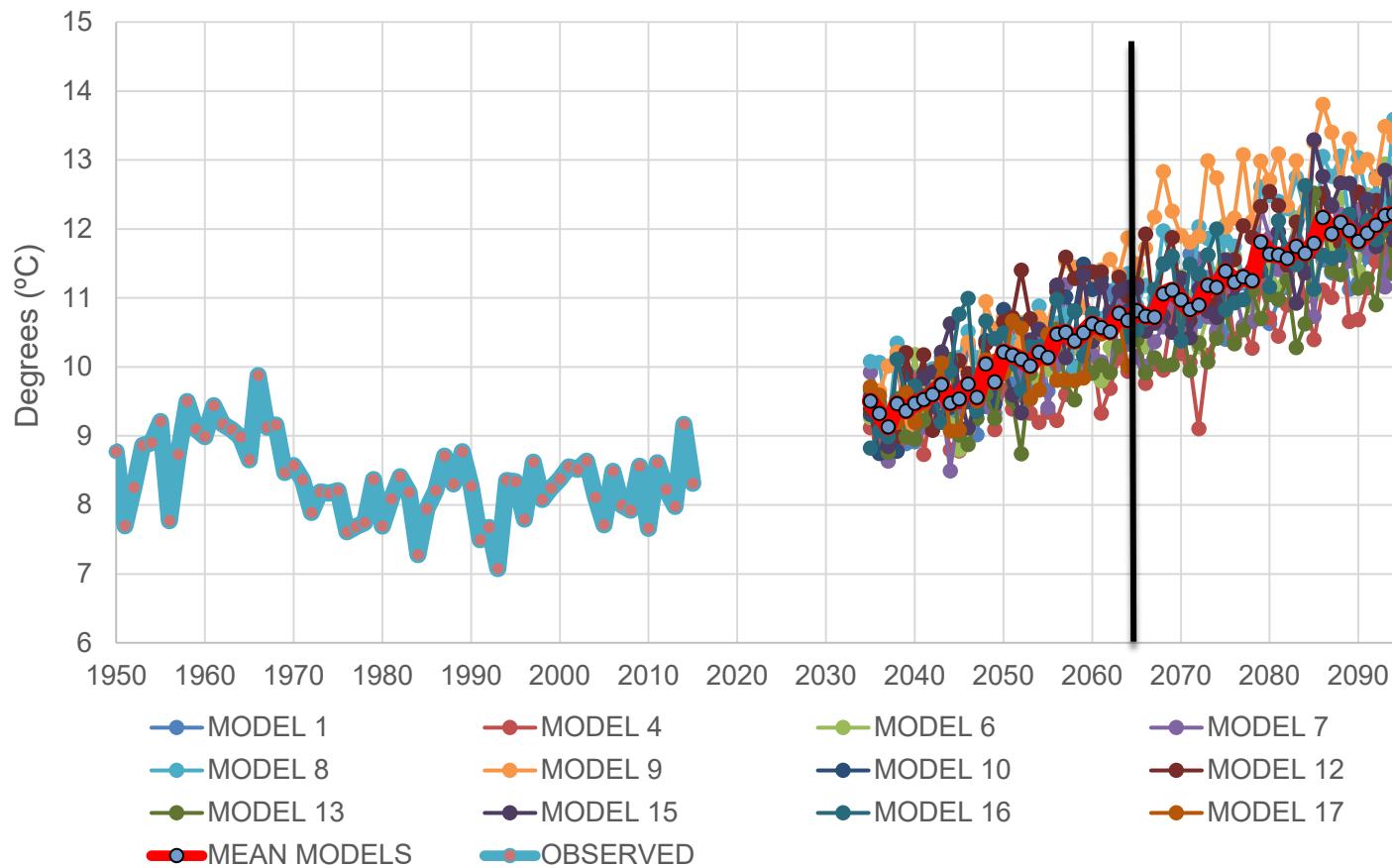
(RCP 8.5)

Model	GCM	RCM	Institute
1	MPI-M-MPI-ESM-LR	COSMO-crCLIM-v1-1	CLMcom-ETH
4	CNRM-CERFACS-CNRM-CM5	CCLM4-8-17	CLMcom
6	CNRM-CERFACS-CNRM-CM5	RACMO22E	KNMI
7	ICHEC-EC-EARTH	COSMO-crCLIM-v1-1	CLMcom-ETH
8	ICHEC-EC-EARTH	RACMO22E	KNMI
9	IPSL-IPSL-CM5A-MR	RACMO22E	KNMI
10	MOHC-HadGEM2-ES	CCLM4-8-17	CLMcom
12	MOHC-HadGEM2-ES	RACMO22E	KNMI
13	MPI-M-MPI-ESM-LR	CCLM4-8-17	CLMcom
15	MPI-M-MPI-ESM-LR	KNMI-RACMO22E	KNMI
16	MPI-M-MPI-ESM-LR	REMO2009	MPI-CSC
17	NCC-NorESM1-M	COSMO-crCLIM-v1-1	CLMcom-ETH

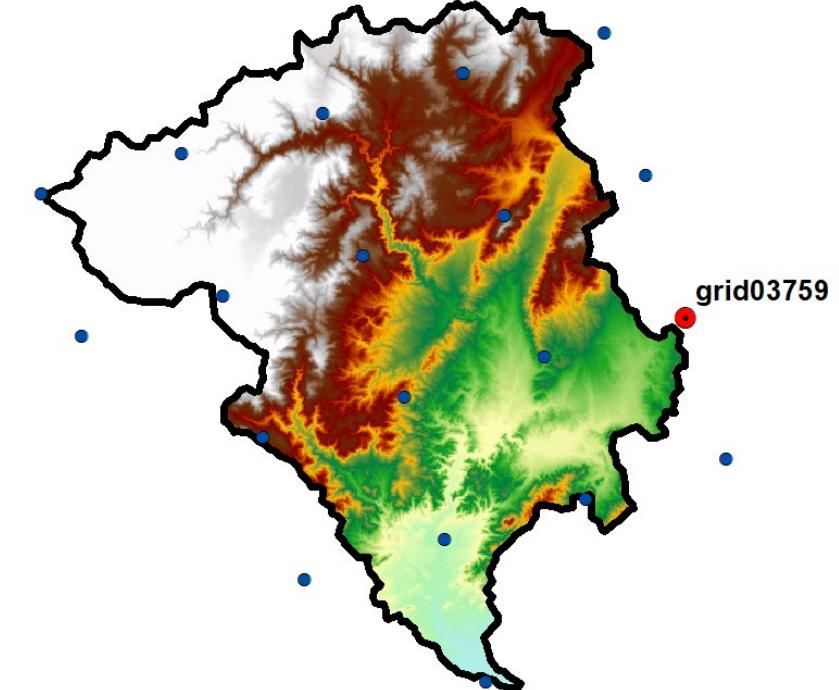
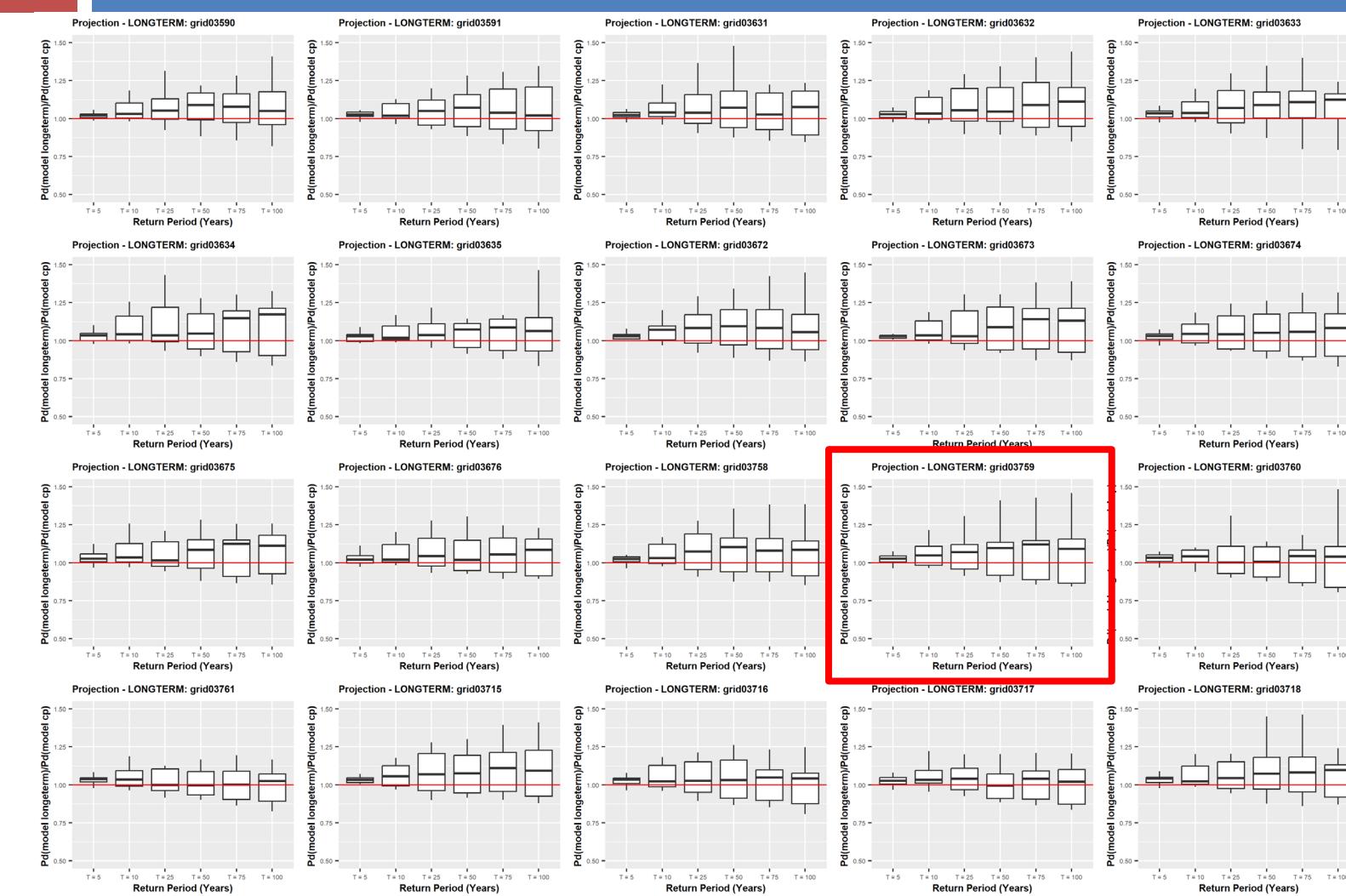
Max. Temperature

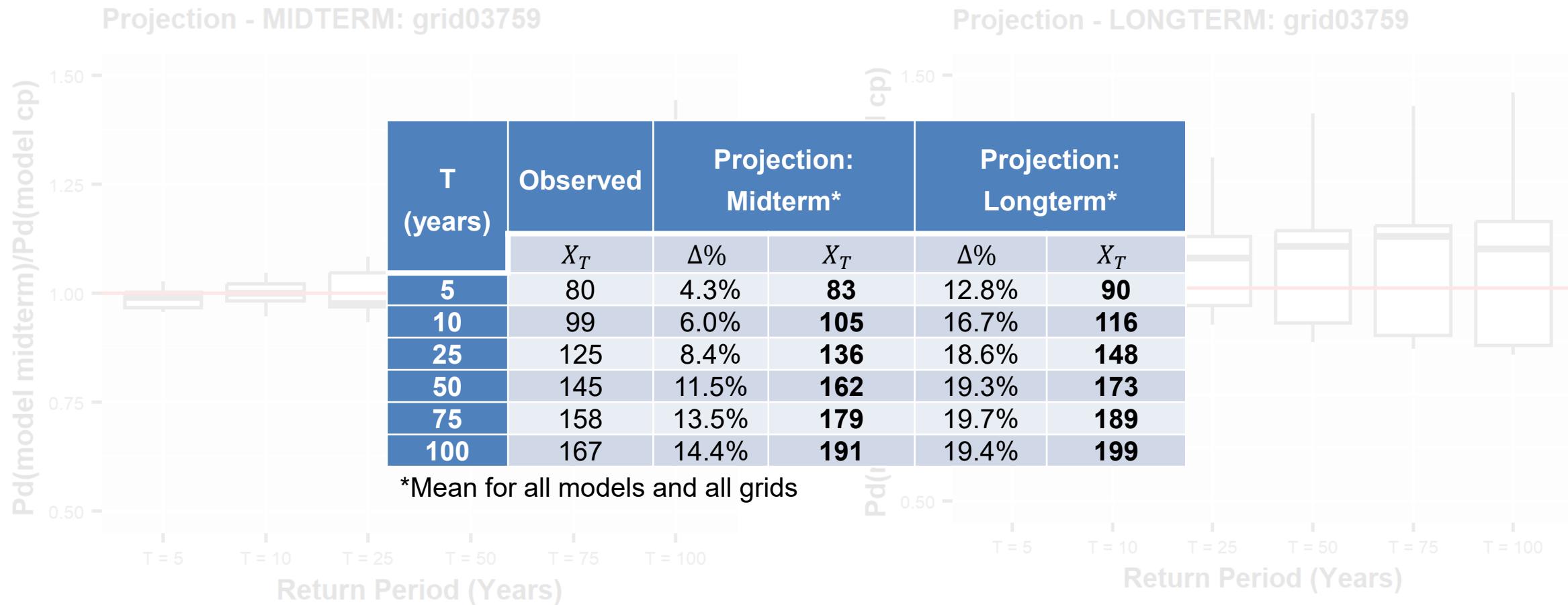


Min. Temperature



Results: Precipitation





Results: Discharges

María Cristina (1,447 km²)

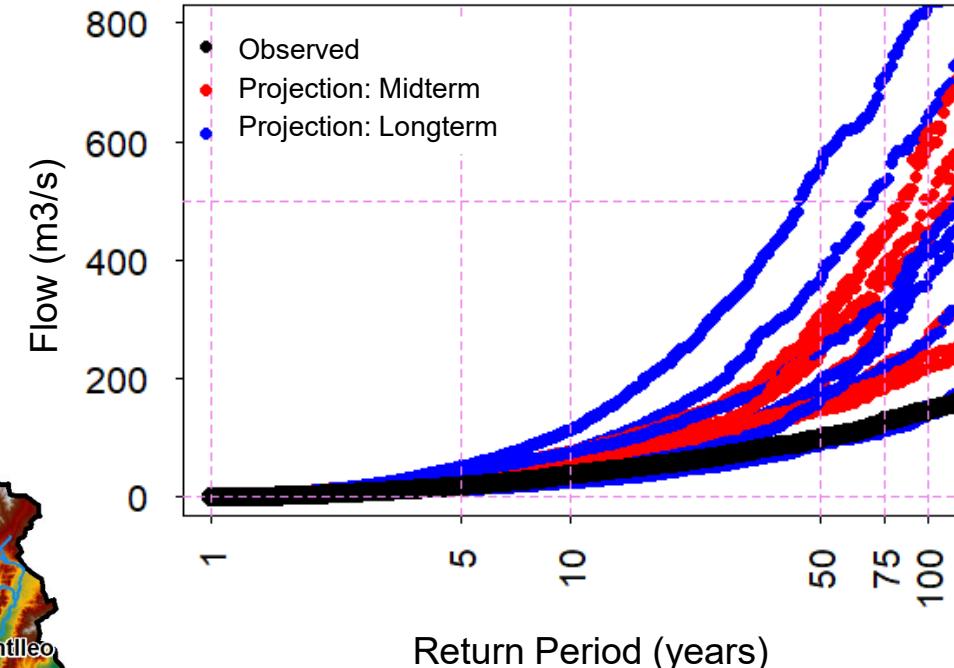
T (years)	OBSERVED (m ³)	CLIMATE PROJECTIONS (m ³)			
		Delta (%)	MIDTERM	Delta (%)	LONGTERM
5	20	12%	22	8%	21
10	38	12%	43	16%	44
25	68	22%	83	33%	91
50	101	38%	140	54%	155
75	130	48%	192	56%	202
100	147	53%	225	58%	232

Vall d'Alba (906 km²)

T (years)	OBSERVED (m ³)	CLIMATE PROJECTIONS (m ³)			
		Delta (%)	MIDTERM	Delta (%)	LONGTERM
5	12	11%	14	10%	13
10	22	13%	24	33%	29
25	39	21%	47	64%	64
50	56	41%	79	88%	105
75	69	49%	103	86%	130
100	80	50%	121	80%	145

Montlleó (501 km²)

T (years)	OBSERVED	CLIMATE PROJECTIONS (m ³)			
		Delta (%)	MIDTERM	Delta (%)	LONGTERM
5	4	3%	4	5%	4
10	6	7%	6	42%	8
25	11	27%	14	111%	23
50	17	57%	27	137%	40
75	21	73%	37	145%	52
100	28	77%	49	130%	64



- The high spatio-temporal variability of floods makes it necessary to use a weather generator in combination with a fully-distributed hydrological model against Design Storm approach and lumped or semi-lumped models.
- Additional information must be incorporated into the implementation of the weather generator for adequate modeling of low frequency quantiles, especially when extreme precipitation records are scarce.
 - Our proposal is to incorporate a regional analysis of annual maximum daily precipitation in climate models to expand the amount of information
- This methodology has been applied in a case study with a semi-arid Mediterranean climate and a reasonable extension, with satisfactory results.

- In terms of temperatures, these will experience a notable increase in both Tmax and Tmin, also increasing the episodes of heat waves and reducing the cold ones.
 - Initial state of the basin: higher ET0
- Precipitation quantiles will also experience a significant increase in both the medium- and long-term projections.
- The methodology estimates a systematic increase in all flood quantiles, which is accentuated in the long-term projection and, especially, when the size of the basin is reduced. Furthermore, these increases in flood quantiles are systematically greater than their respective precipitation quantiles.



Thank you for your attention!



1973



1982



2012



2019

Research Institute of Water and Environmental Engineering
Universitat Politècnica de València

This study has been supported by the Ministry of Science, Innovation, and Universities of Spain through the Research Projects TETISCHANGE (RTI2018-093717-B-100) and TETISPREDICT (PID2022-141631OB-I00).