

# Downscaling of ECMWF seasonal integrations by RegCM

Čedo Branković and Mirta Patarčić

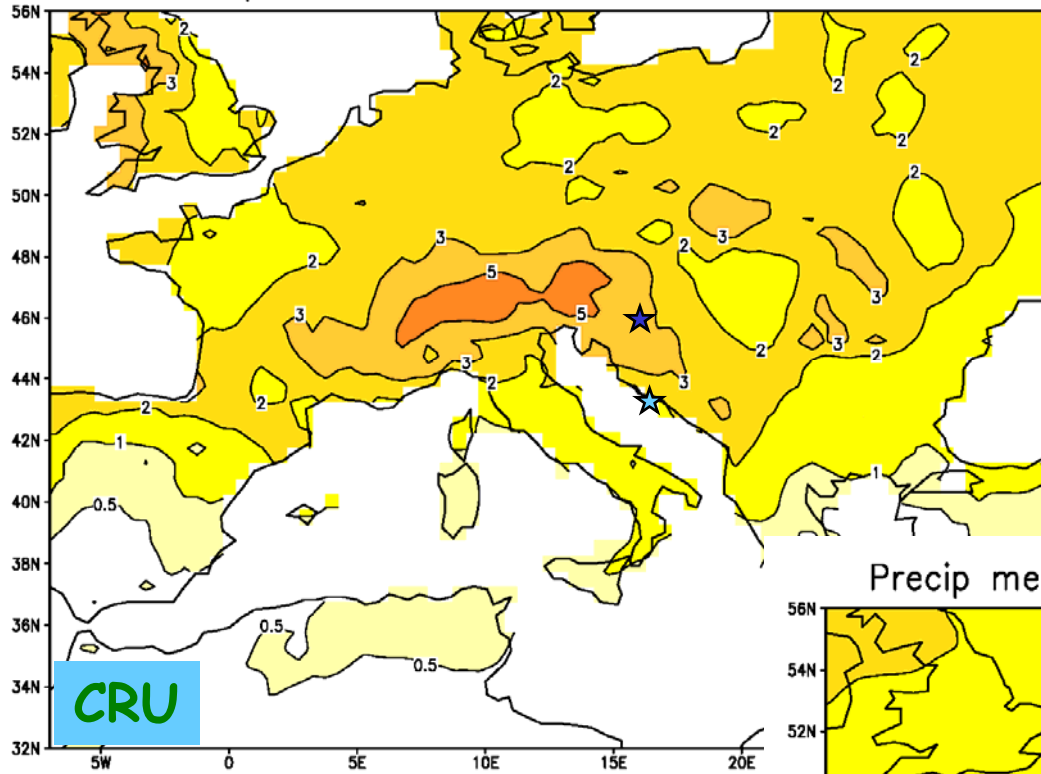
Croatian Meteorological and Hydrological Service  
Grič 3, 10000 Zagreb, Croatia

(Thanks to Paul Dando and Manuel Fuentes for help in data retrieval)

## Outline of the talk

1. Introduction
2. Downscaling of ECMWF operational seasonal forecasts (an attempt)
3. Downscaling of ECWMF experimental seasonal forecasts (ENSEMBLES)
4. Some conclusions

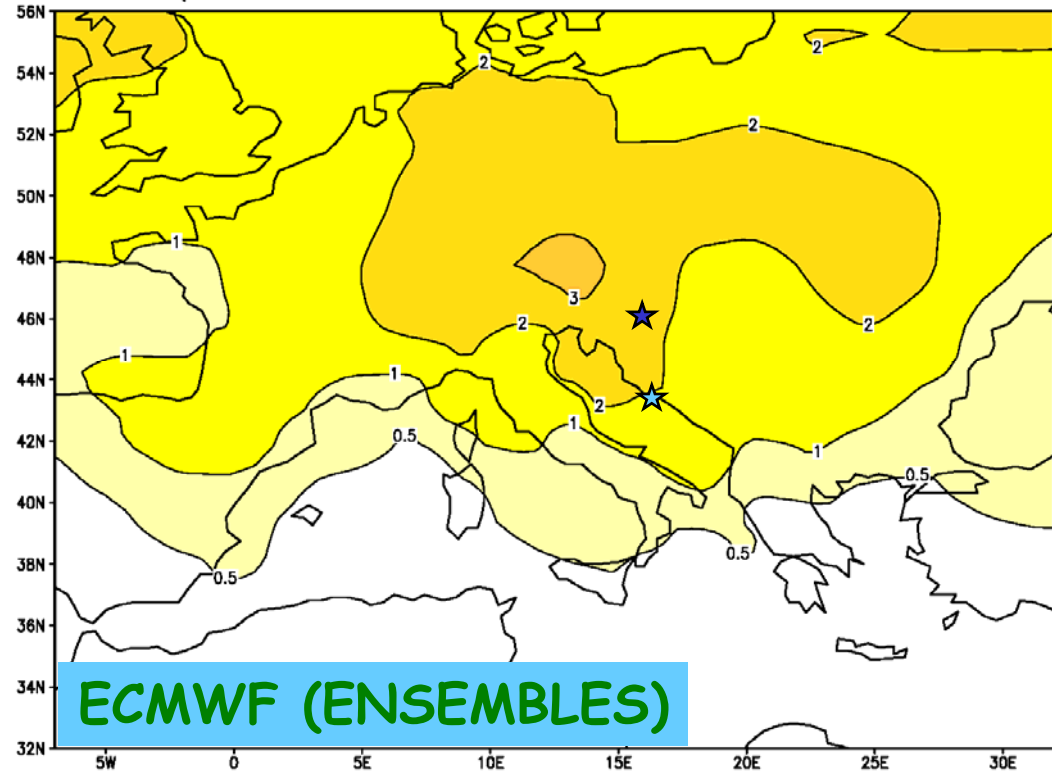
Precip mean JAS; CRU; cint=0.5 1 2 3 5 10



Precipitation (mm/day)

JAS  
(mean 1991-2001)

Precip mean JAS; ECSF; cint=0.5 1 2 3 5 10 15 20

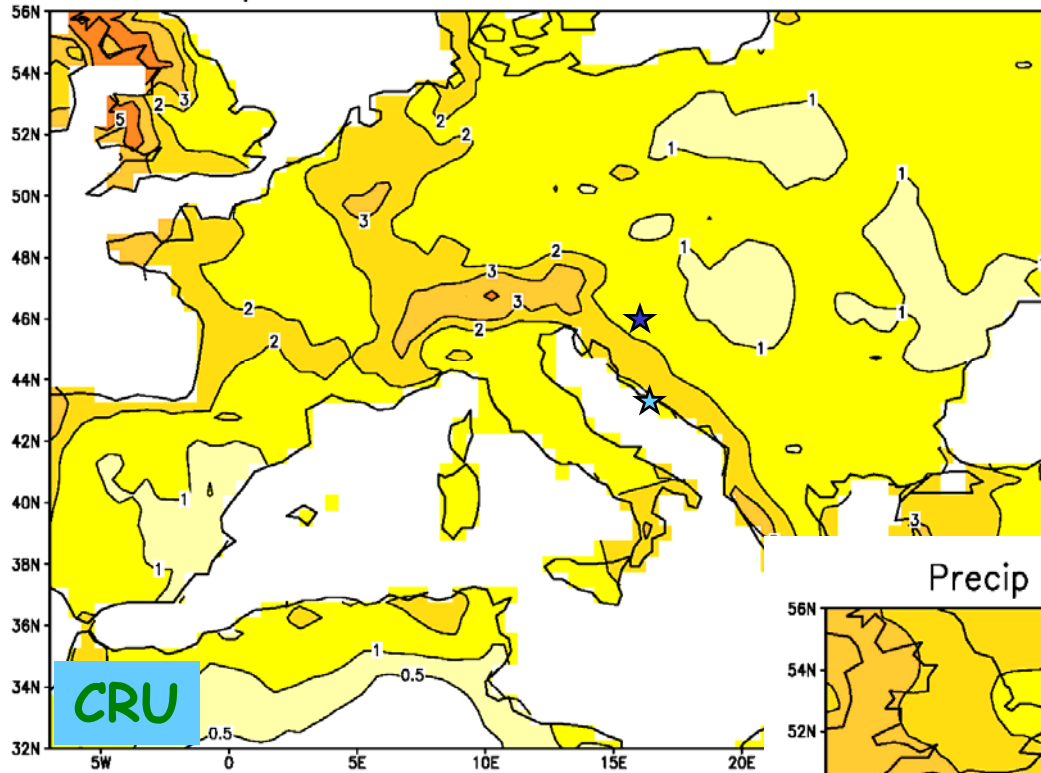


Observed (mm/day)

Zagreb 2.9 ★

Split 1.5 ★

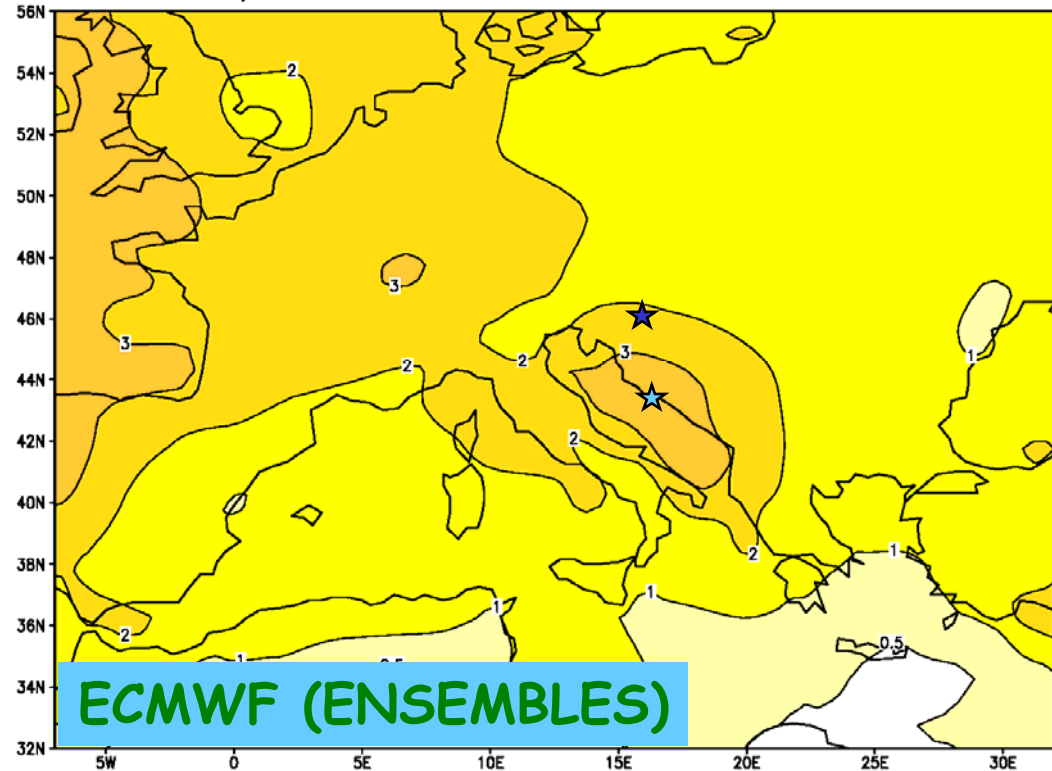
Precip mean JFM; CRU; cint=0.5 1 2 3 5 10



Precipitation (mm/day)

JFM  
(mean 1992-2002)

Precip mean JFM; ECSF; cint=0.5 1 2 3 5 10

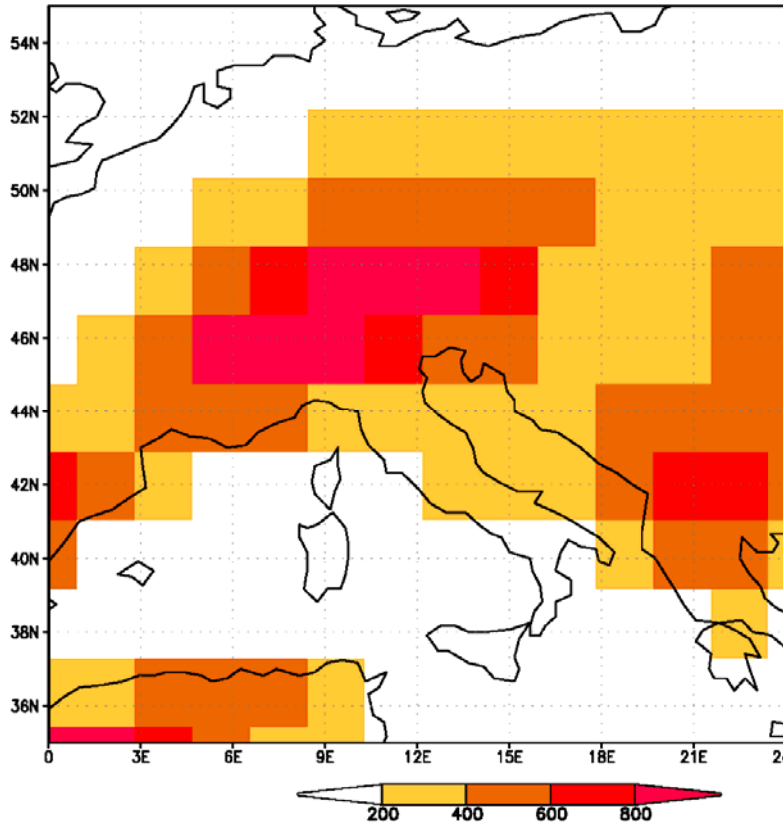


Observed (mm/day)

Zagreb 1.3 ★

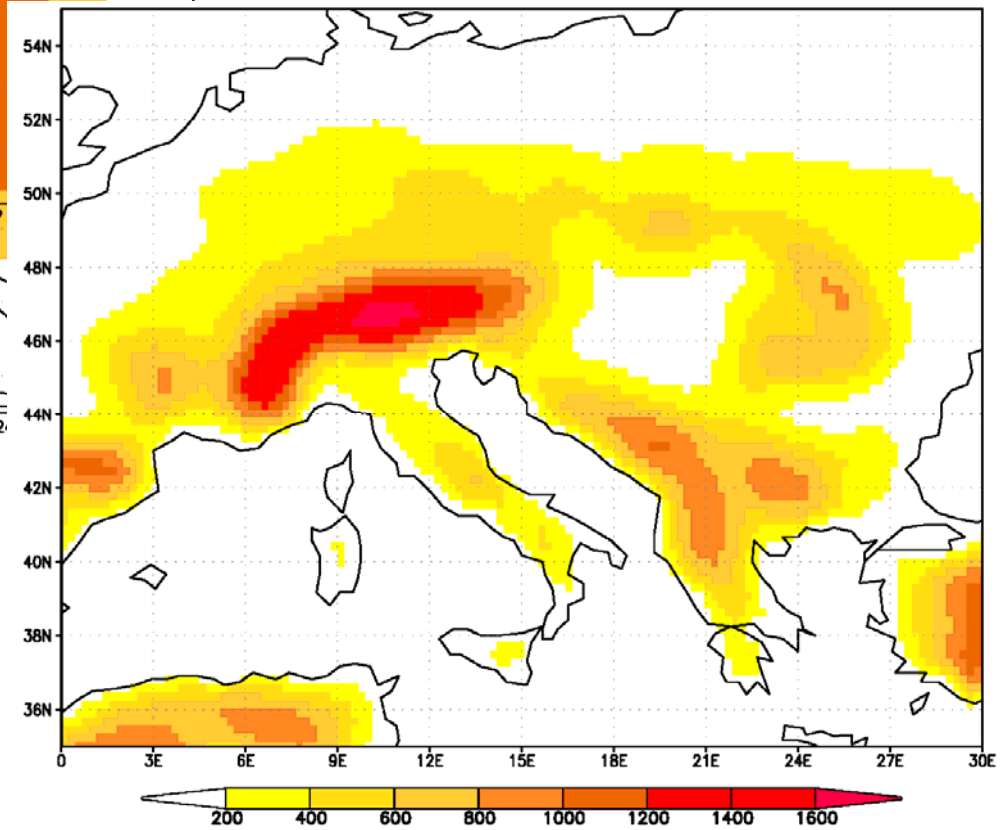
Split 1.7 ★

# Orography in global and limited area models



ECMWF 1.875°

RegCM 50 km



## Outline of the talk

1. Introduction
2. Downscaling of ECMWF **operational** seasonal forecasts (an attempt)
3. Downscaling of ECWMF experimental seasonal forecasts
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## ECMWF operational seasonal forecasts

- \* T<sub>L</sub>95 global spectral model (1.875°)
- \* 40 members, JJA 2003 (only one season tested)
- \* 6 pressure levels (1000, 925, 850, 700, 500, 200)
- \* Frequency of LBCs every 12 hours

Big ensemble, but poor input for a viable downscaling.

Do we have to downscale the whole ensemble?

RegCM (Giorgi et al. 1993, MWR): 50 km, 14 levels

## How to select a sub-ensemble?

- \* From short- and medium-range forecasting experience:

Objectively select **representative members** that characterise all possible evolution scenarios of the global model ensemble.

(Molteni et al. 2001 *QJ*, Montani et al. 2001 *Nonlin. Proc. Geophys.*)

- \* If resources allow, **downscale all members** and then manipulate

Representative members from a global model may not be representative in a regional model.

(Experiments at CMHS: downscaling of ECMWF EPS by using Aladin-HR; Branković et al. 2007 ECMWF Tech Memo 507)



# Z500

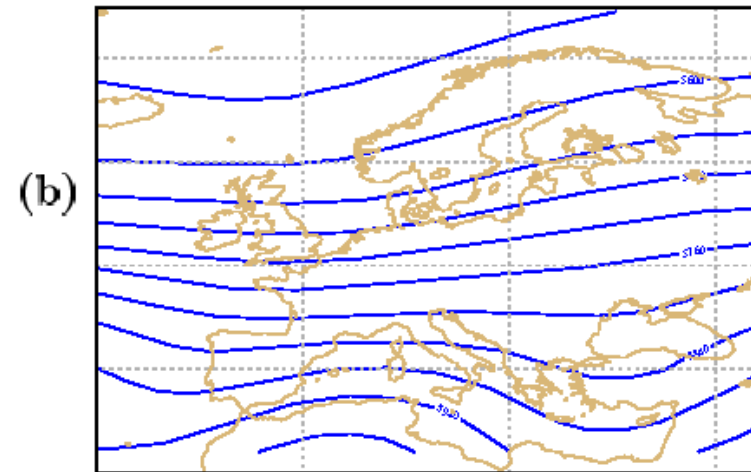
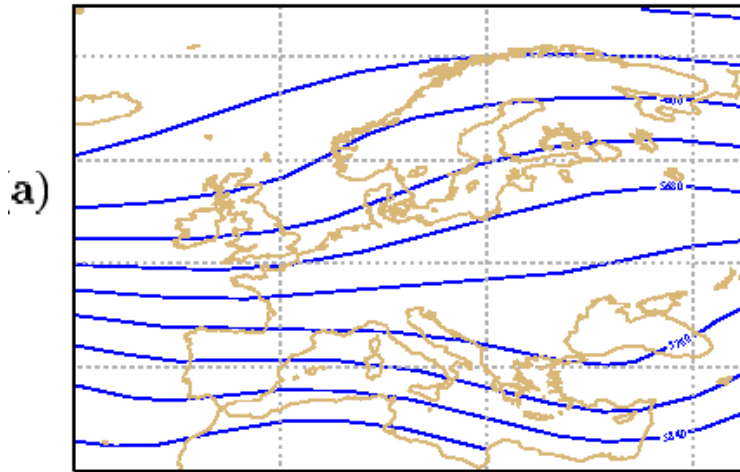
June 2003

Most populated clusters

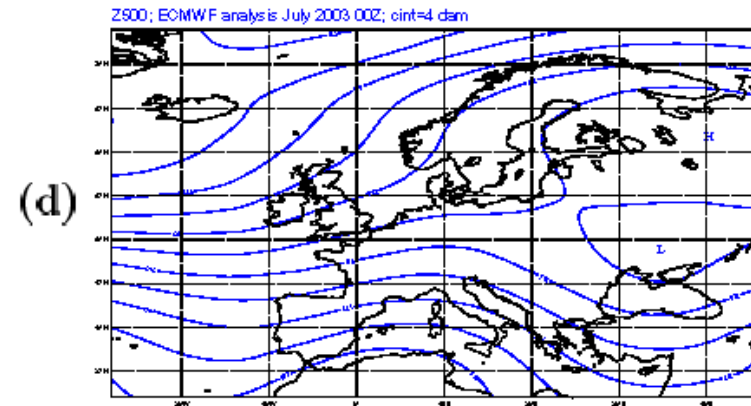
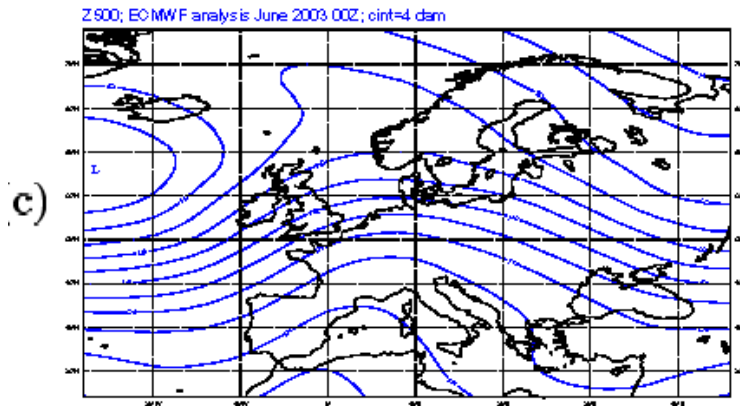
July 2003

20030501+2 mon z500 cluster 2 - 18 memb  
Clustering. area: 75N -20W 30S 45E

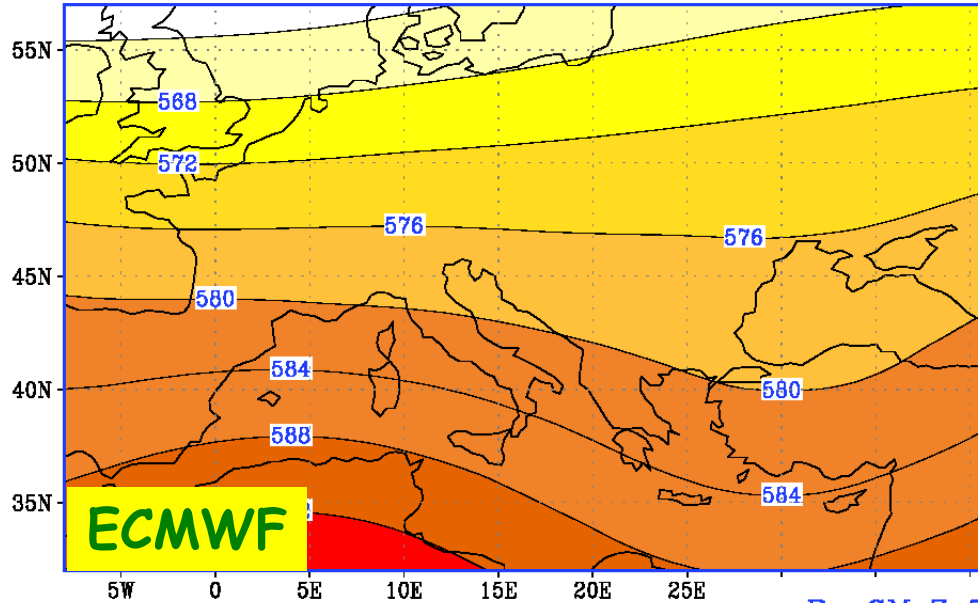
20030501+3 mon z500 cluster 1 - 24 memb  
Clustering. area: 75N -20W 30S 45E



ECMWF operational analysis



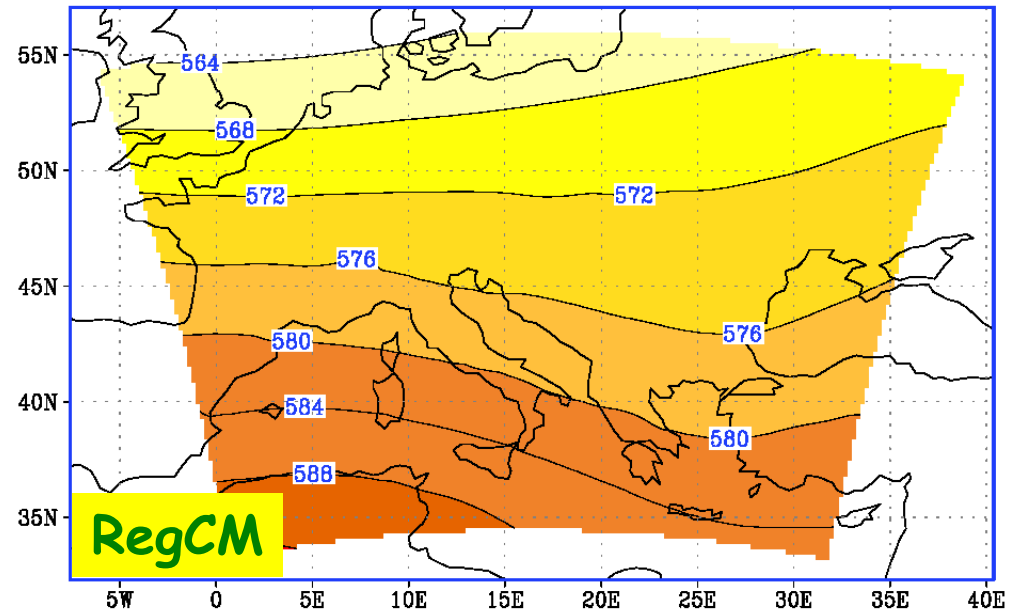
ECMWF Z 500; JJA 2003 (12hr); ens. mean (12)  
cont=4 dam



# Z500 JJA 2003

Ensemble mean  
(12 members)

RegCM Z 500; JJA 2003 (12hr); ensemble mean (12)  
cont=4 dam

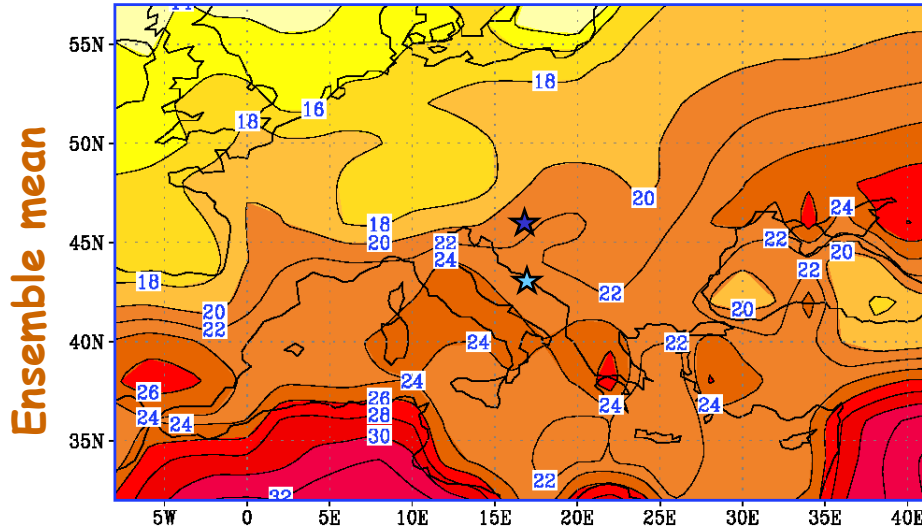


# ECMWF

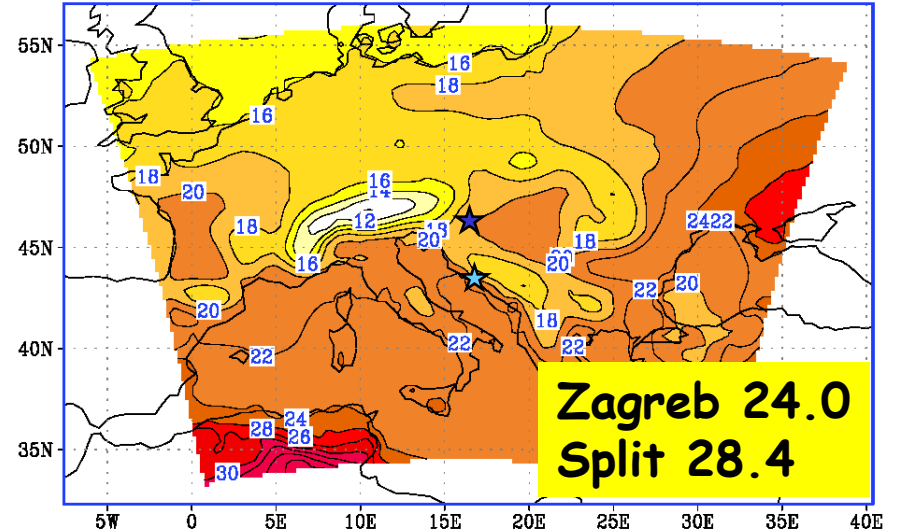
# T2m JJA 2003

# RegCM

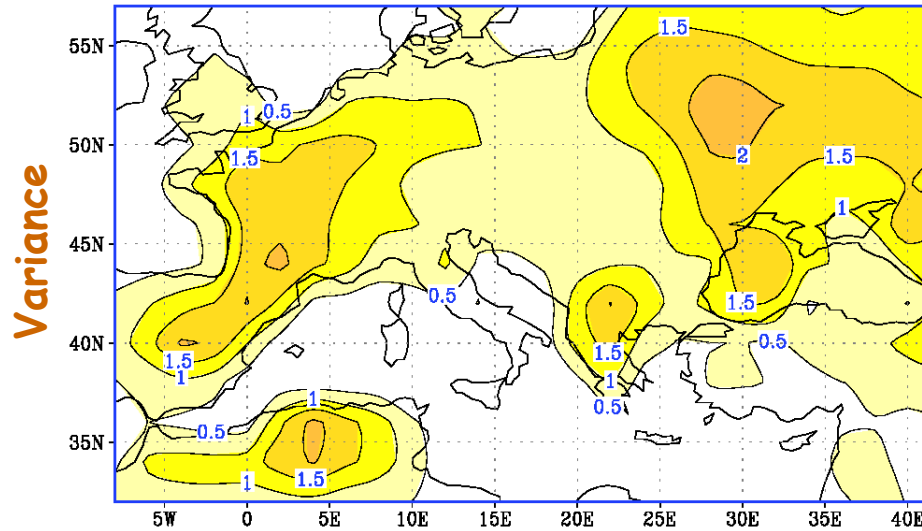
ECMWF T2m; JJA 2003; ensemble mean (6hr)  
cont=2 deg



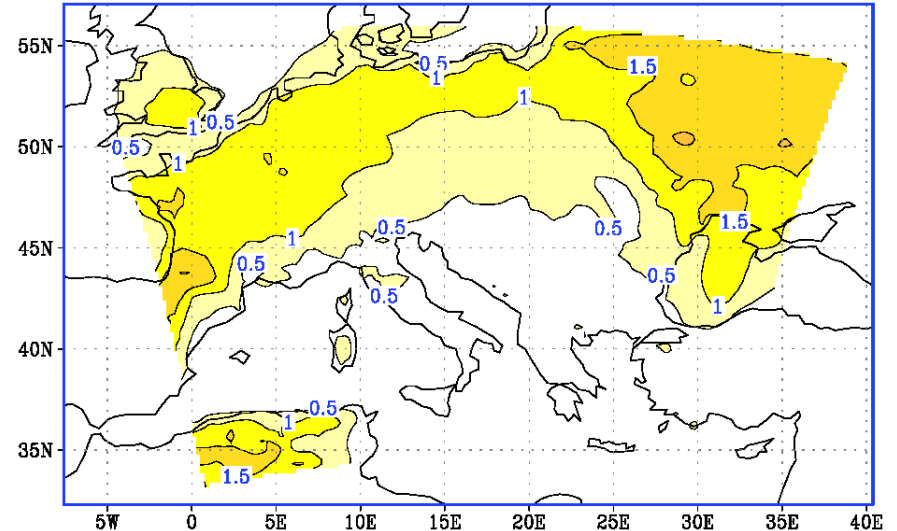
RegCM T2m; JJA 2003 (allZ); ens. mean (12)  
cont=2 deg



ECMWF T2m; JJA 2003; variance (6hr)  
cont=0.5 deg



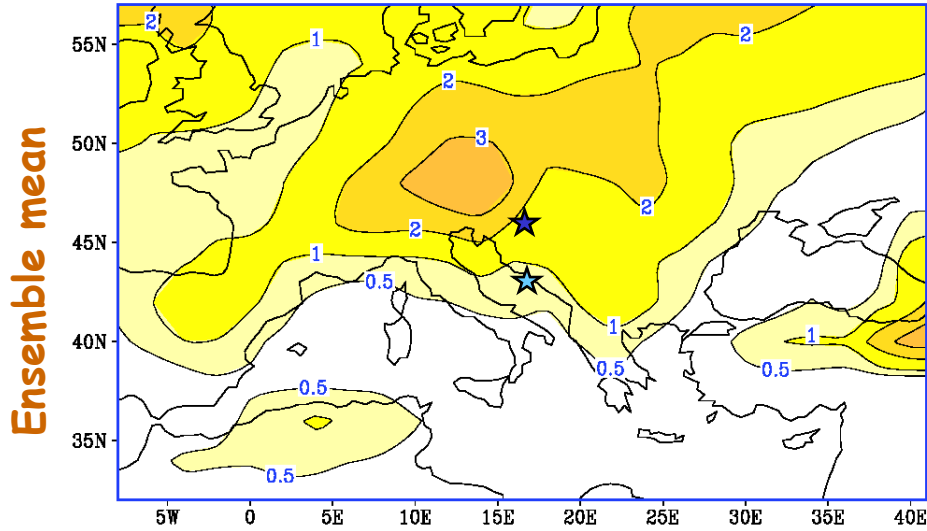
RegCM T2m; JJA 2003 (allZ); variance (12)  
cont=0.5 1 1.5 2 2.5 3 3.5 4 4.5 5



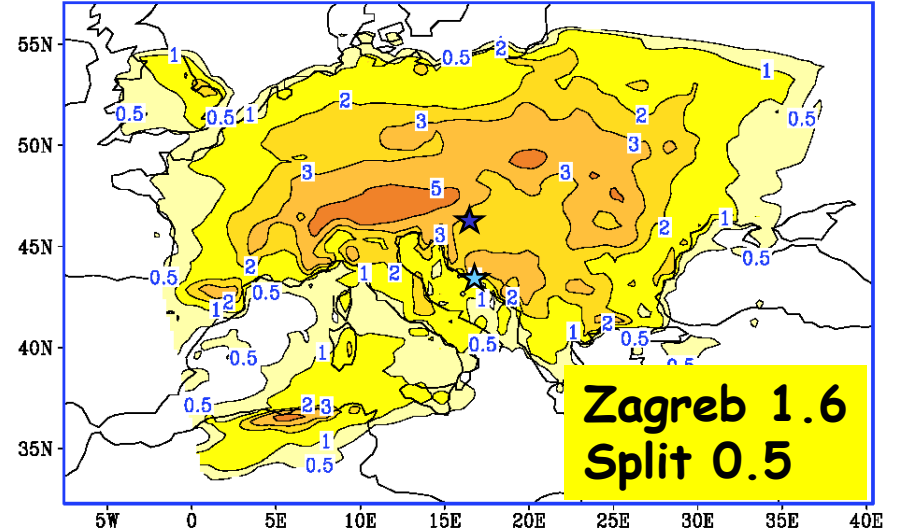
# ECMWF Precipitation JJA 2003

RegCM

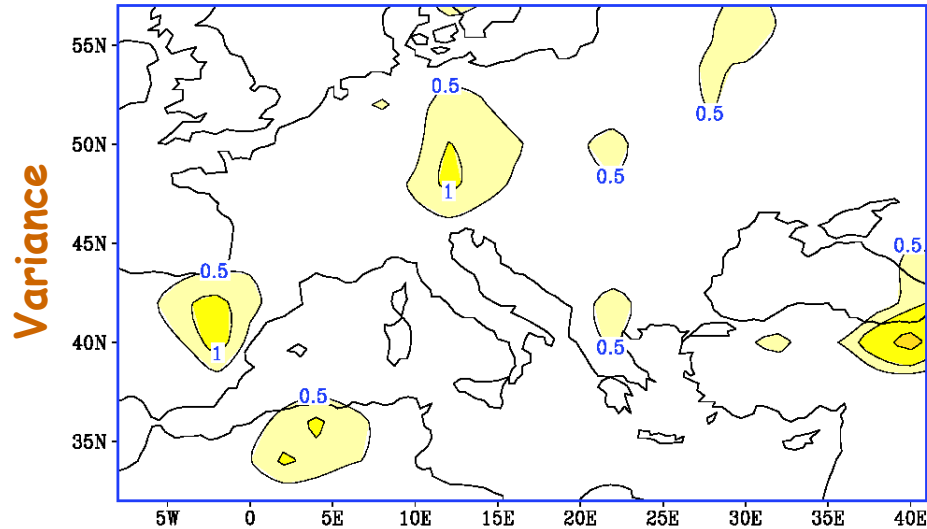
ECMWF Total precip; JJA 2003; ensemble mean  
cont=0.5 1 2 3 5 10 15 20 30 mm/day



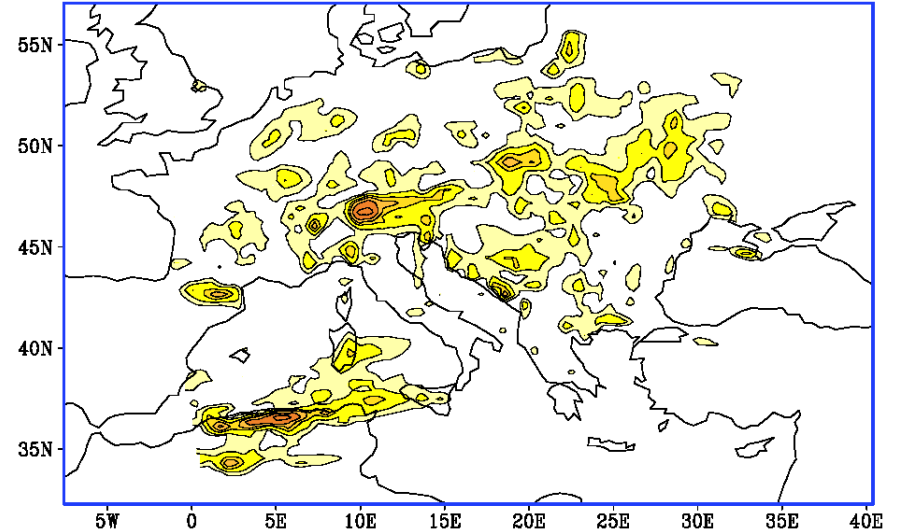
RegCM Total precip; JJA 2003; ensemble mean (allZ)  
cont=0.5 1 2 3 5 10 15 20 30 mm/day



ECMWF Total precip; JJA 2003; variance  
cont=0.5 1 2 3 5 10 15 20 30 50 mm/day



RegCM Total precip; JJA 2003; variance (allZ)  
cont=0.5 1 2 3 5 10 15 20 30 50



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1. Introduction
2. Downscaling of ECMWF operational seasonal forecasts (an attempt)
3. Downscaling of ECMWF **experimental** seasonal forecasts
4. Some conclusions

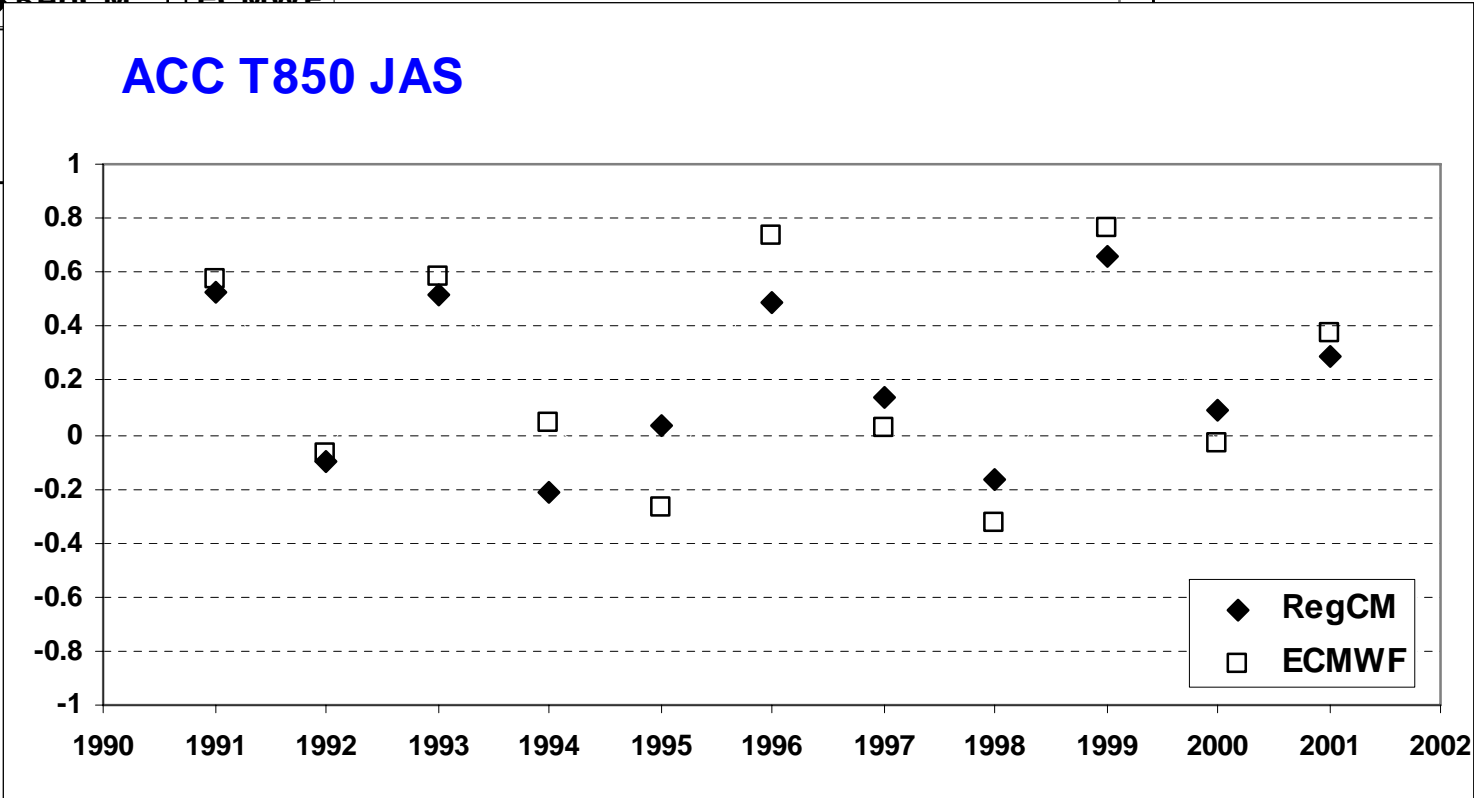
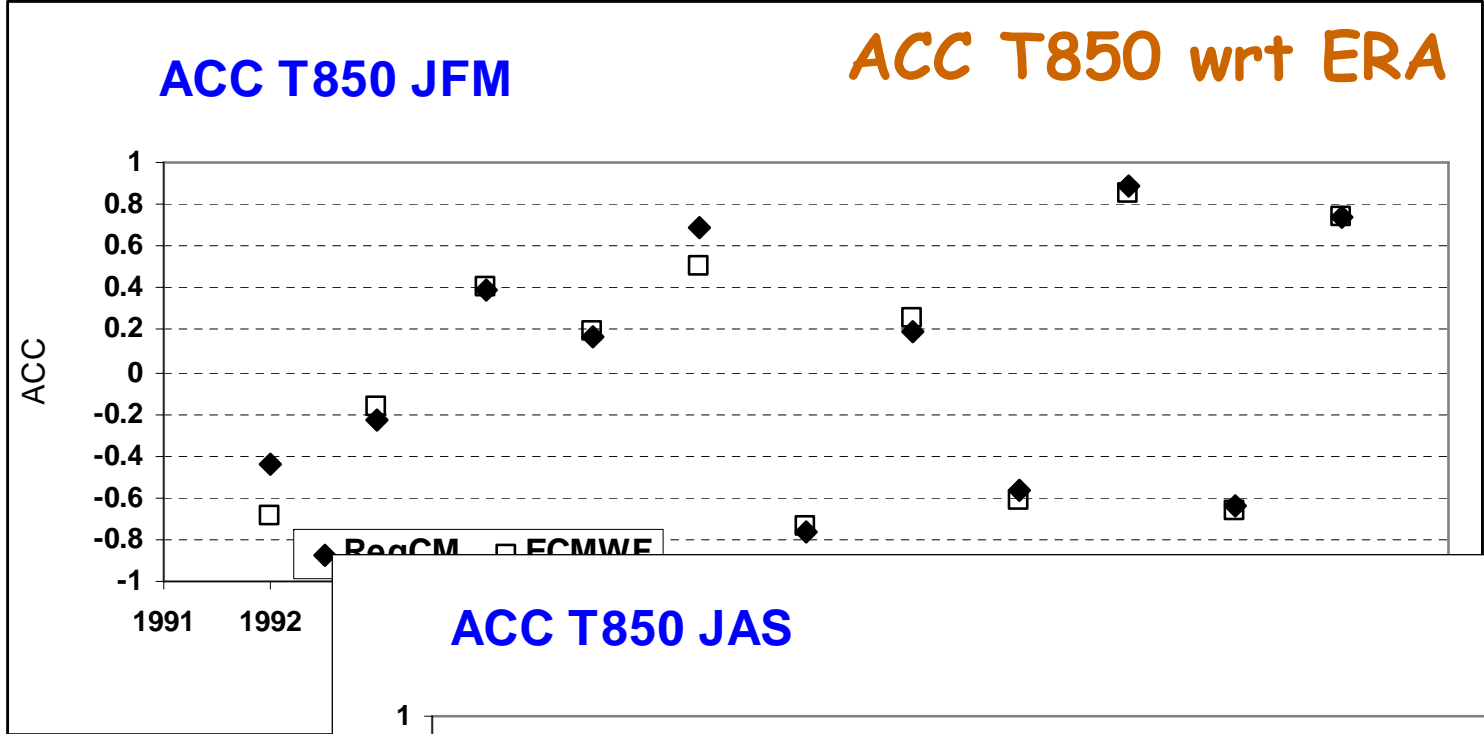
## ECMWF experimental seasonal forecasts

- \* Part of ENSEMBLES project
- \* T<sub>L</sub>95 global spectral model (1.875°), 40 model levels
- \* Frequency of LBCs every 6 hours
- \* 6-month f/c (May, November), 9 members, 1991-2001

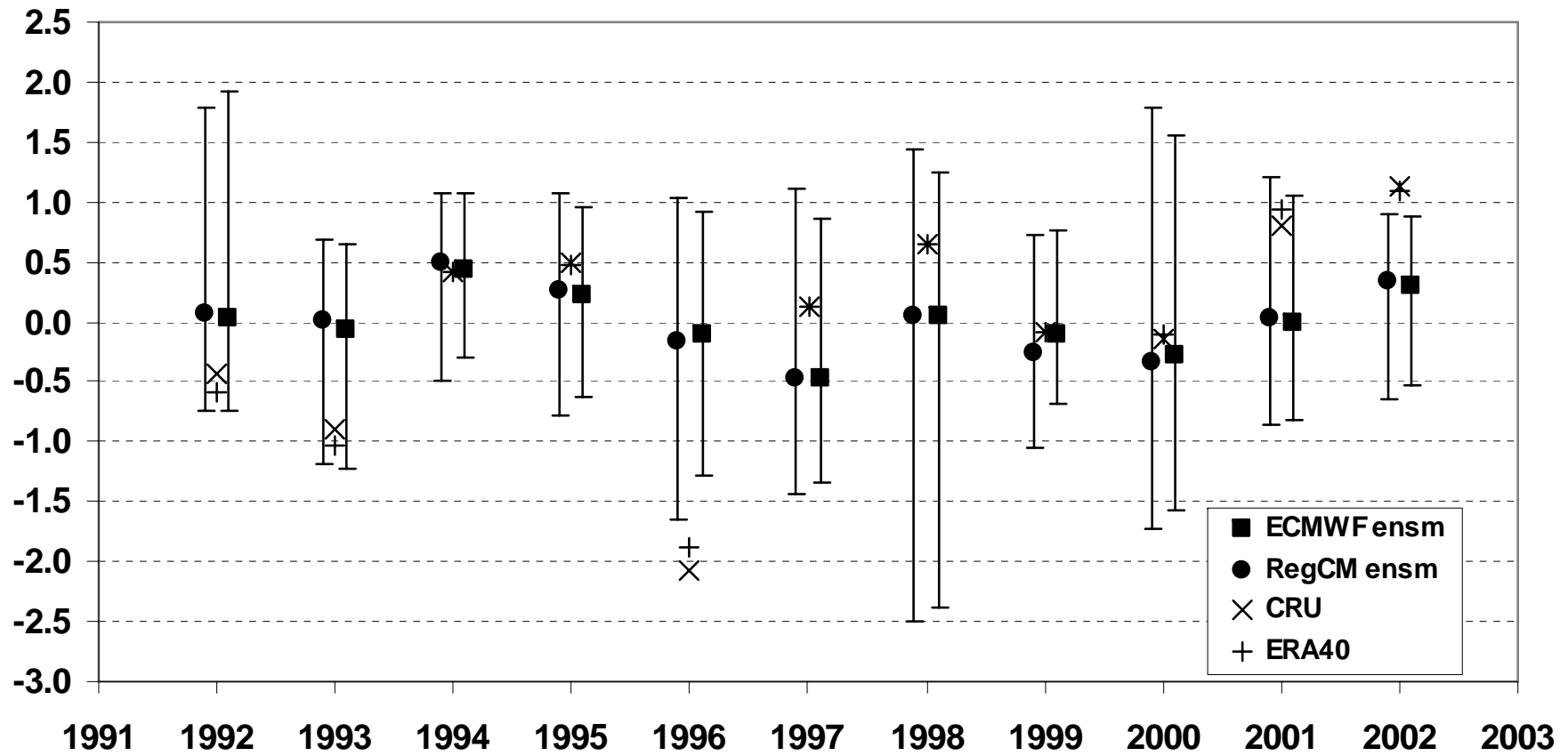
Much better input for downscaling, but smaller ensembles.

No sub-ensembles!

RegCM: 50 km, 18 levels; JAS and JFM seasons

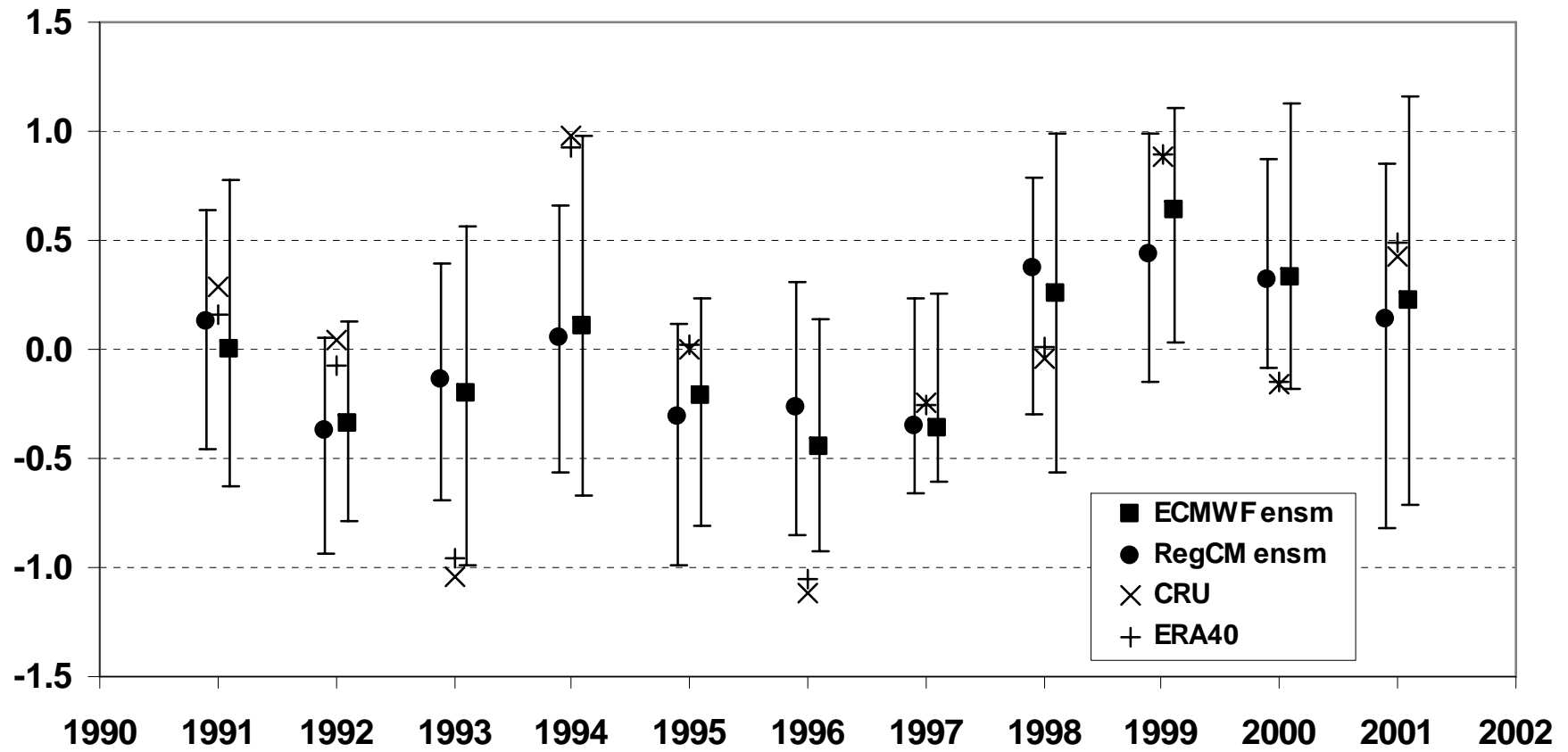


## T2m anomaly JFM





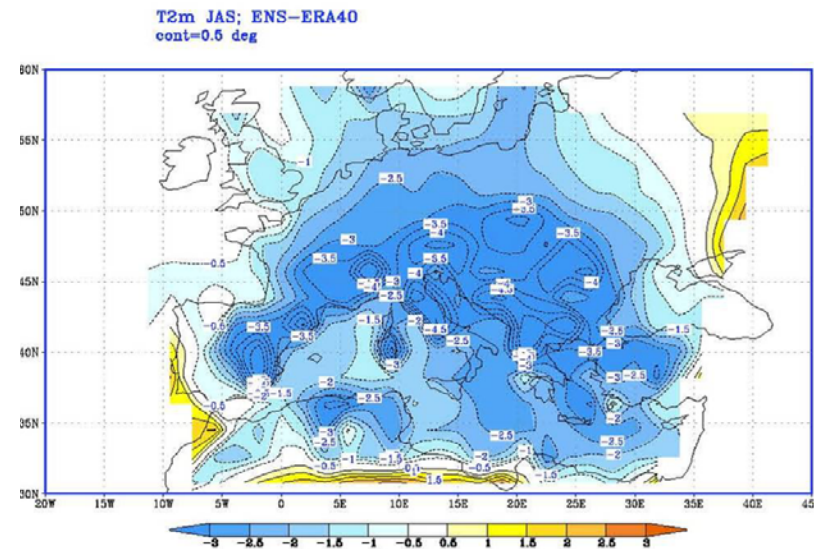
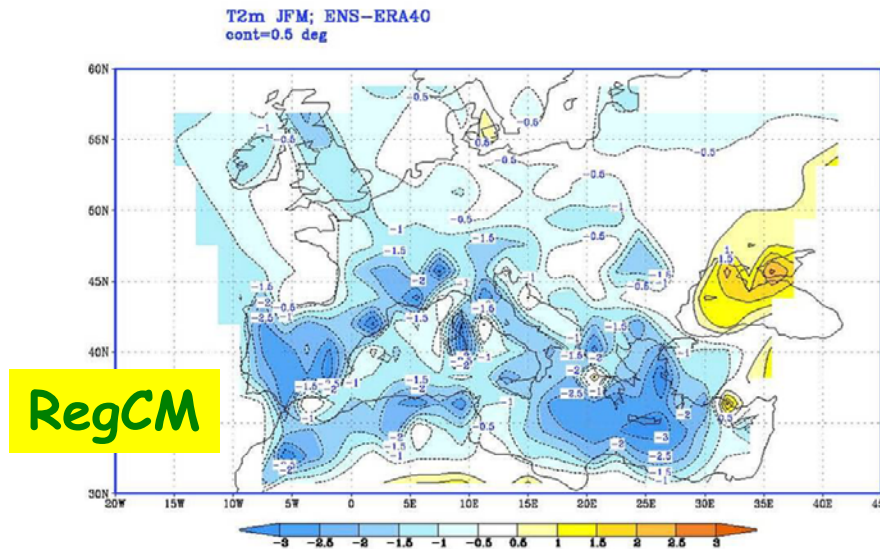
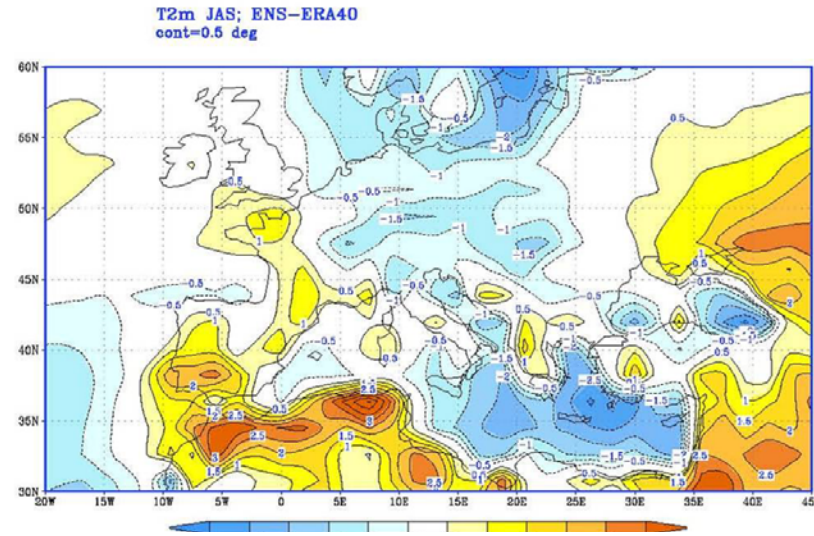
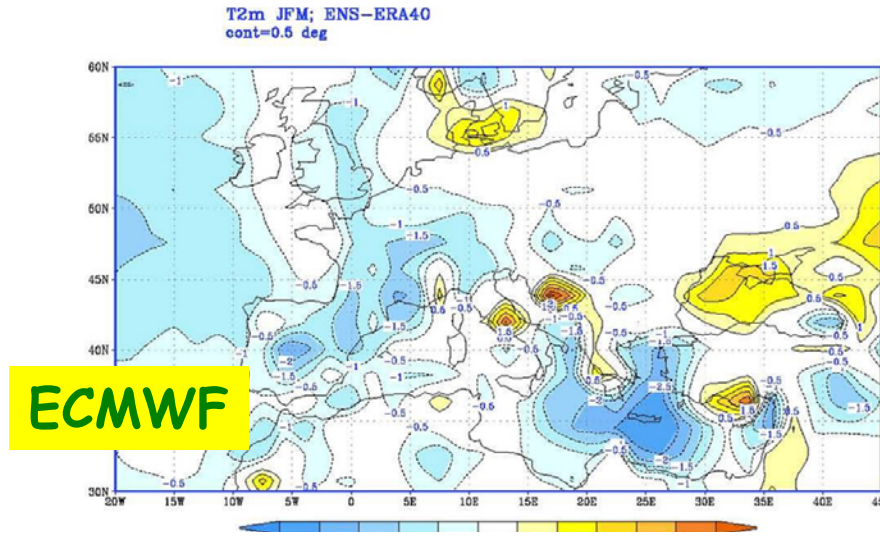
## T2m anomaly JAS



# T2m error (11 years)

## JFM

## JAS

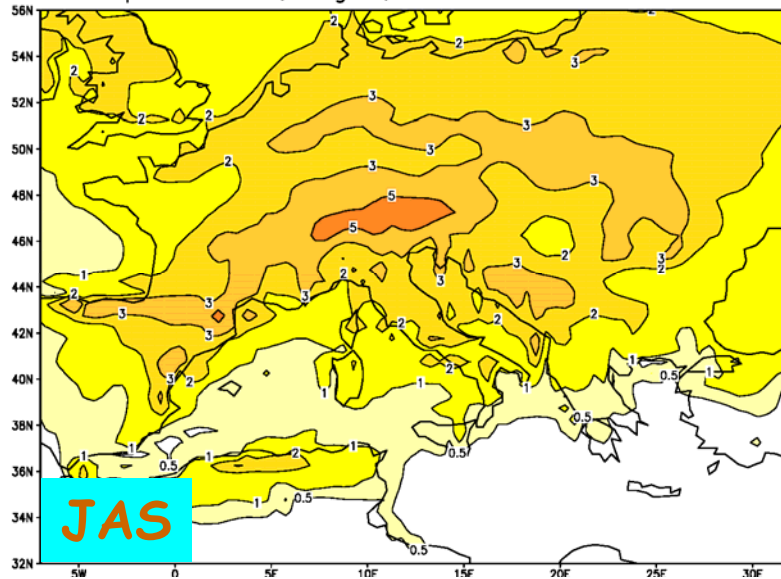


RegCM

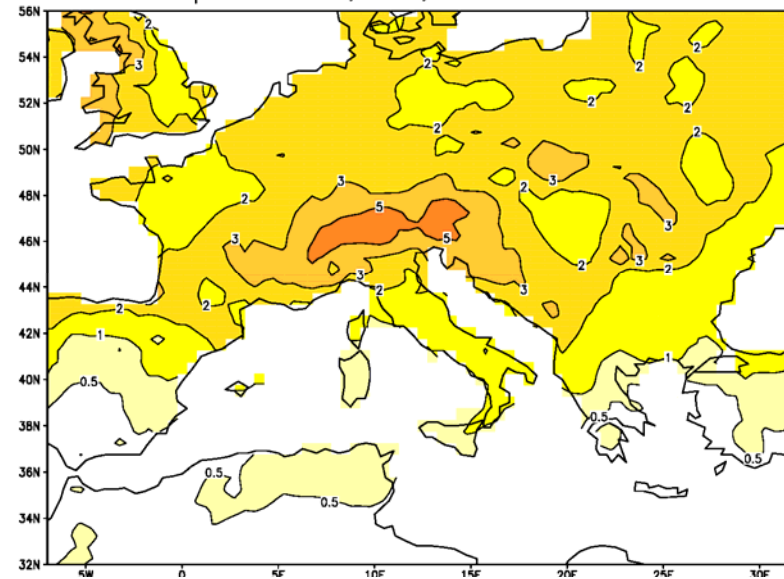
# Precipitation (11 years)

CRU verif

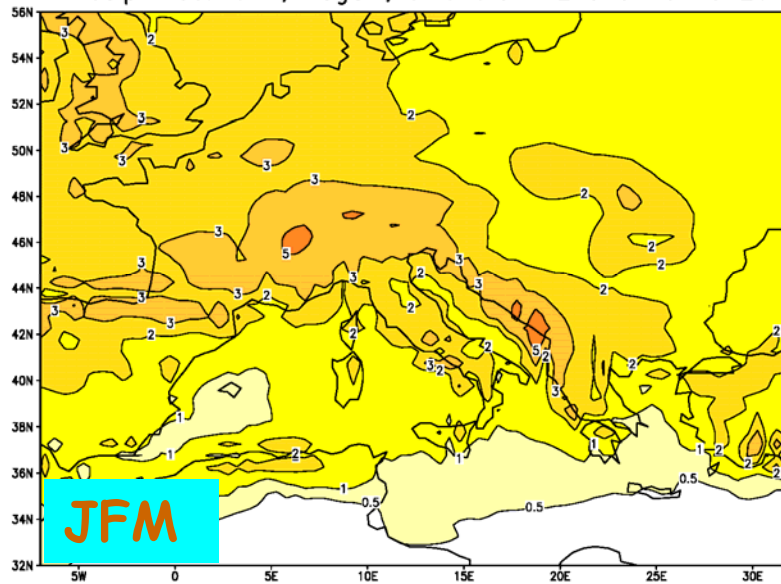
Precip mean JAS; RegCM; cint=0.5 1 2 3 5 10 15 20



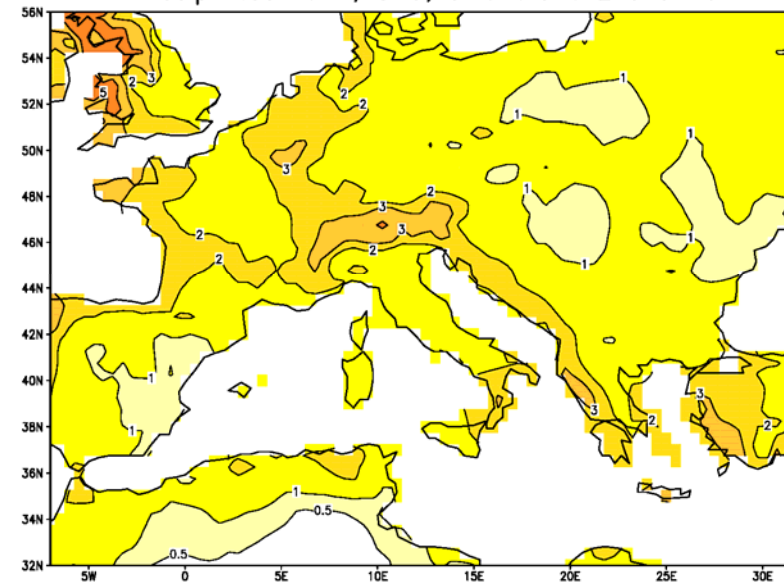
Precip mean JAS; CRU; cint=0.5 1 2 3 5 10



Precip mean JFM; RegCM; cint=0.5 1 2 3 5 10 15 20



Precip mean JFM; CRU; cint=0.5 1 2 3 5 10



## Skill scores (accuracy measures)

Contingency tables and quantities (Wilks 1995)

		Observed	
		Y	N
Fcst	Y	a	b
	N	c	d

$n = a+b+c+d$  total number of fcst/event pairs

Heidke score [1 perfect, 0 random f/c,  $HSS < 0$  worse than random f/c]

$$HSS = \frac{2(ad-bc)}{(a+c)(c+d)+(a+b)(b+d)}$$

$$HR = \frac{a+d}{n} \quad \text{Hit rate [0,1]}$$

$$FAR = \frac{b}{a+b} \quad \text{False alarm ratio [1,0]}$$

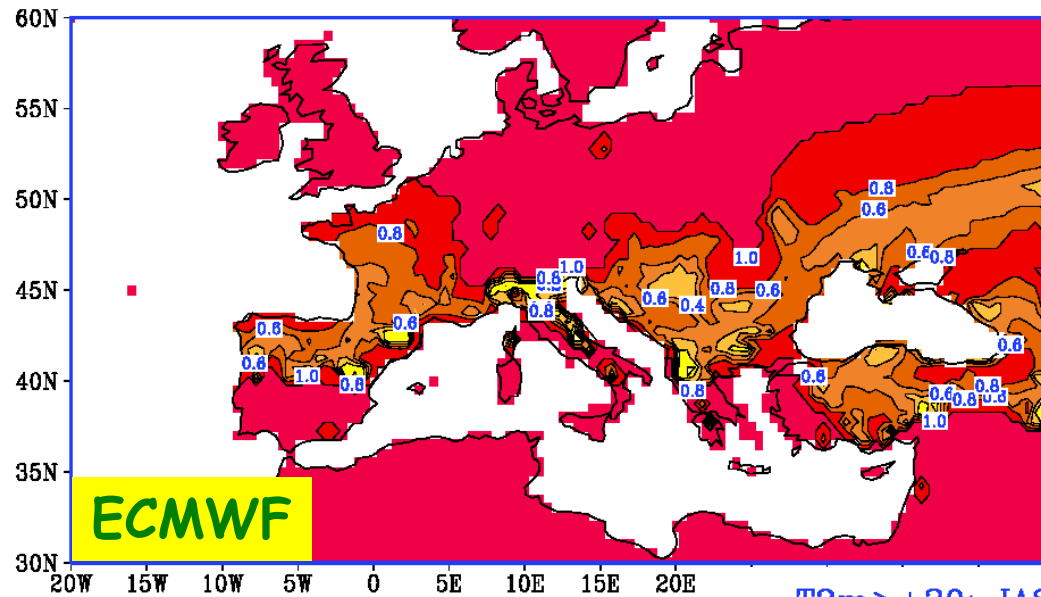
$$TS = \frac{a}{a+b+c} \quad \text{Threat score [0,1]}$$

$$\text{Kuipers score } KSS = \frac{ad-bc}{(a+c)(b+d)}$$

Bias [1 unbiased,  $B > 1$  overforecasting,  $B < 1$  underforecasting]

$$B = \frac{a+b}{a+c}$$

T2m>+20; JAS Hit rate; ECMWF (CRU)  
cont=0.001 0.2 0.4 0.6 0.8 0.999



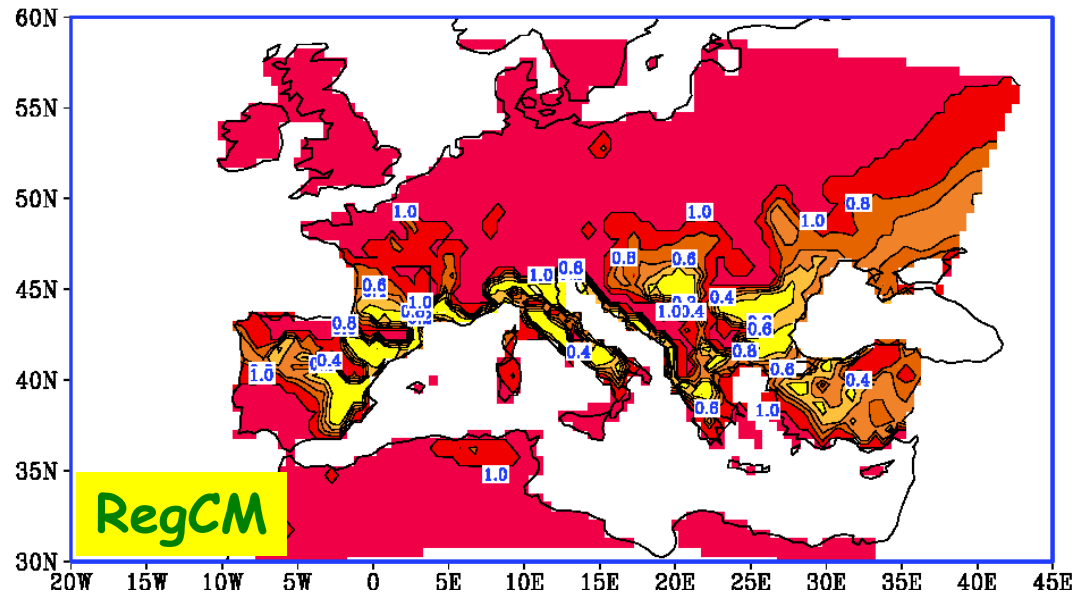
T2m > +20°C  
JAS

Hit rate (wrt CRU)

$$HR = \frac{a+d}{n}$$

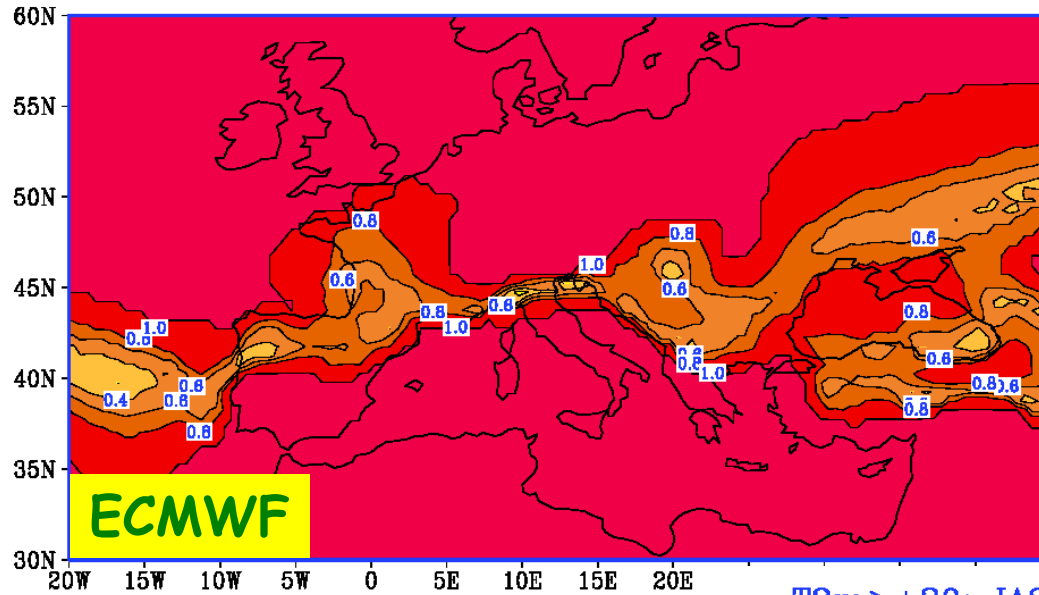
		Observed	
		Y	N
Fcst	Y	a	b
	N	c	d

T2m>+20; JAS Hit rate; RegCM (CRU)  
cont=0.001 0.2 0.4 0.6 0.8 0.999





T2m > +20; JAS Hit rate; ECMWF (ERA)  
 cont=0.001 0.2 0.4 0.6 0.8 0.999



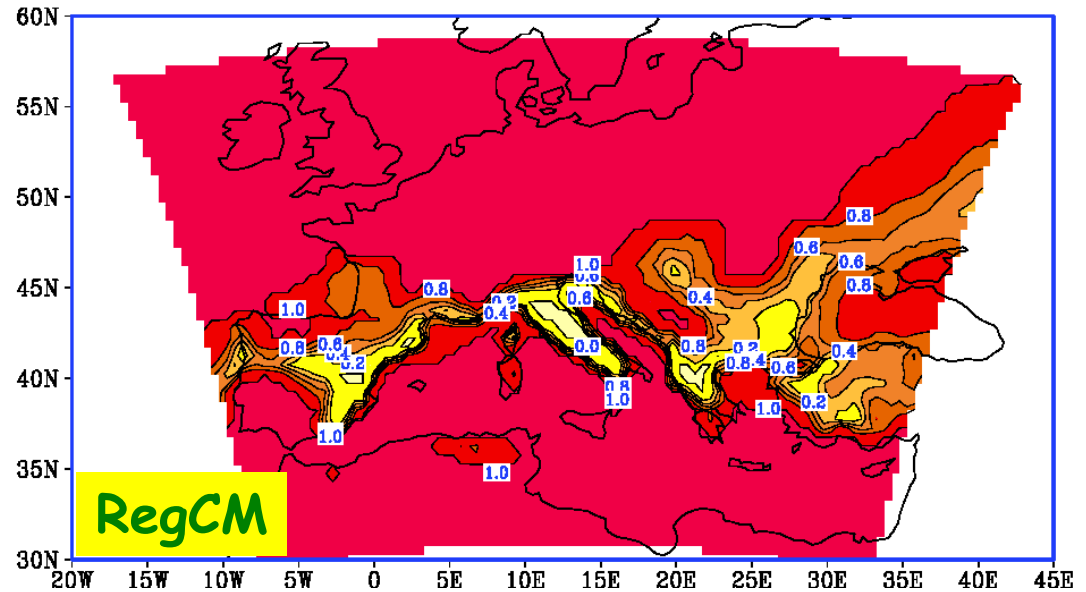
T2m > +20°C  
 JAS

Hit rate (wrt ERA)

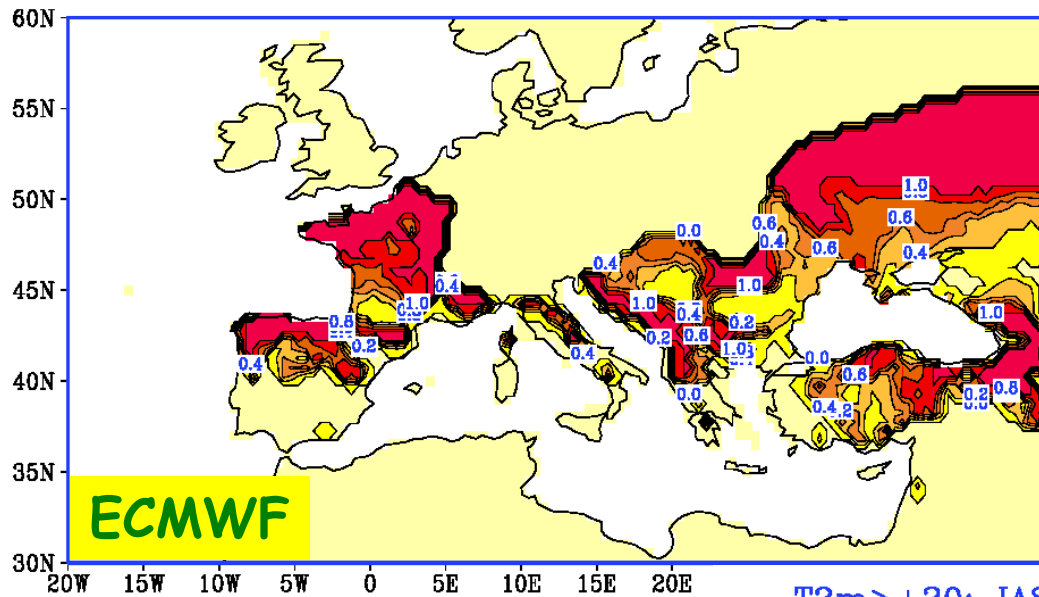
$$HR = \frac{a+d}{n}$$

		Observed	
		Y	N
Fcst	Y	a	b
	N	c	d

T2m > +20; JAS Hit rate; RegCM (ERA)  
 cont=0.001 0.2 0.4 0.6 0.8 0.999

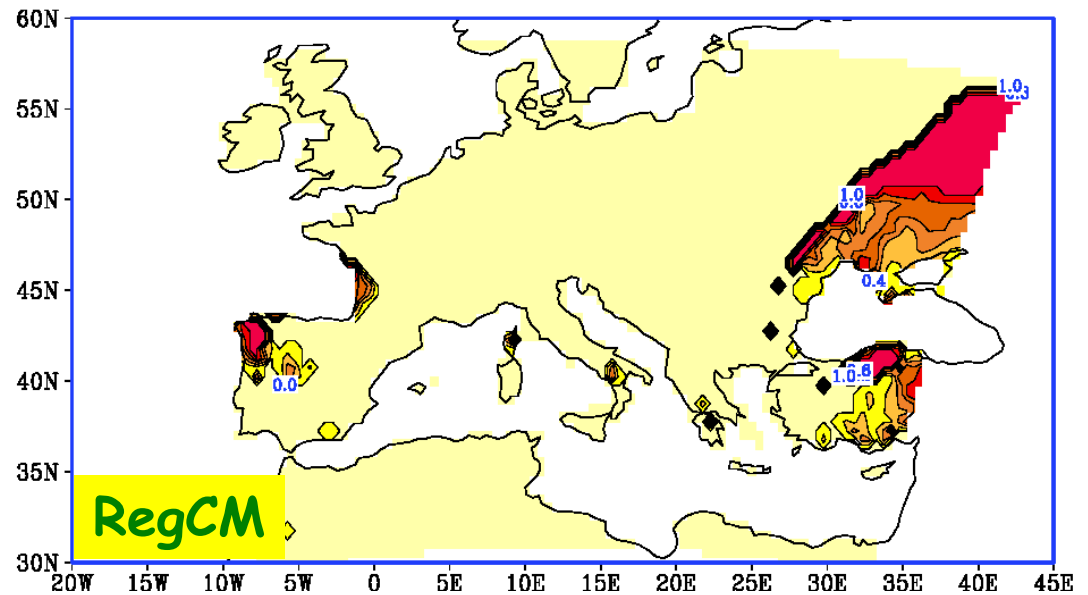


T2m > +20; JAS False alarm rate; ECMWF (CRU)  
 cont=0.001 0.2 0.4 0.6 0.8 0.999



T2m > +20°C  
 JAS  
 False alarm ratio

T2m > +20; JAS False alarm rate; RegCM (CRU)  
 cont=0.001 0.2 0.4 0.6 0.8 0.999



$$FAR = \frac{b}{a+b}$$

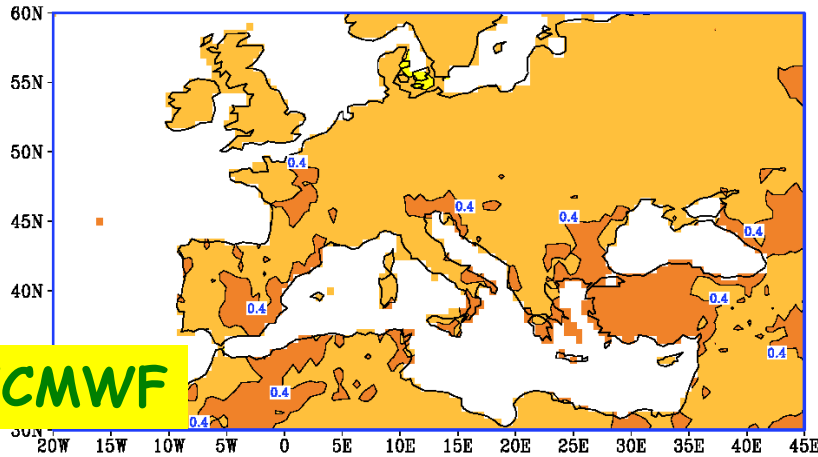
		Observed	
		Y	N
Fcst	Y	a	b
	N	c	d

T2m anom > +0°C Threat score  $TS = \frac{a}{a+b+c}$

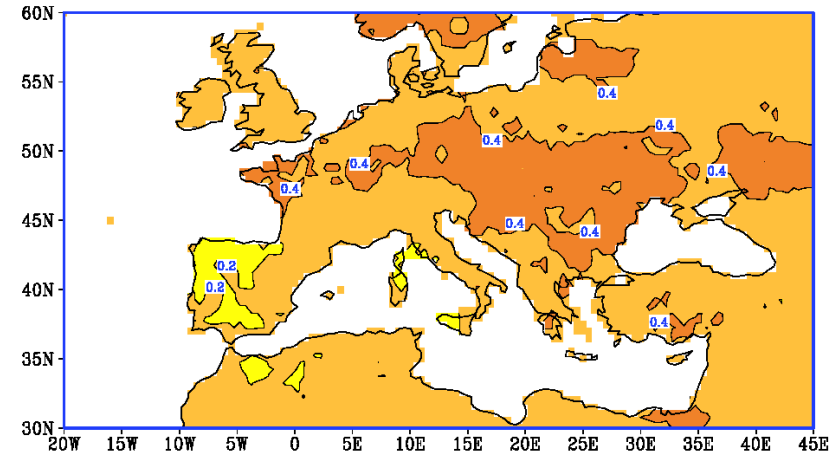
JAS

JFM

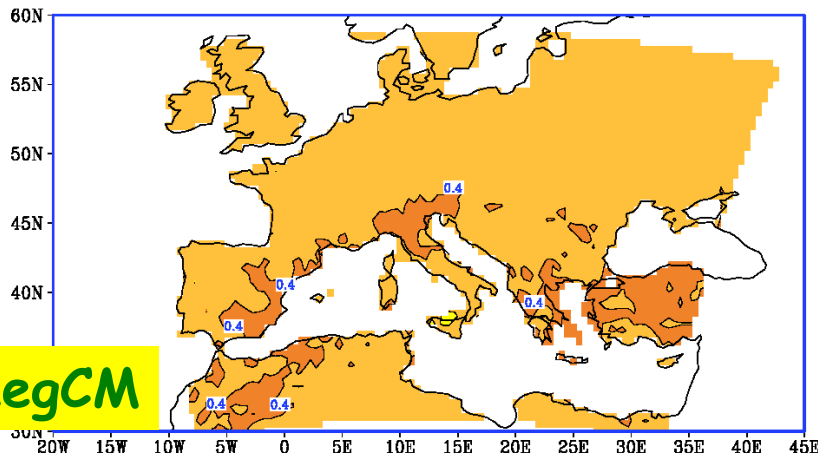
T2m anom>+0; JAS Threat score; ECMWF (CRU)  
cont=0.001 0.2 0.4 0.6 0.8 0.999



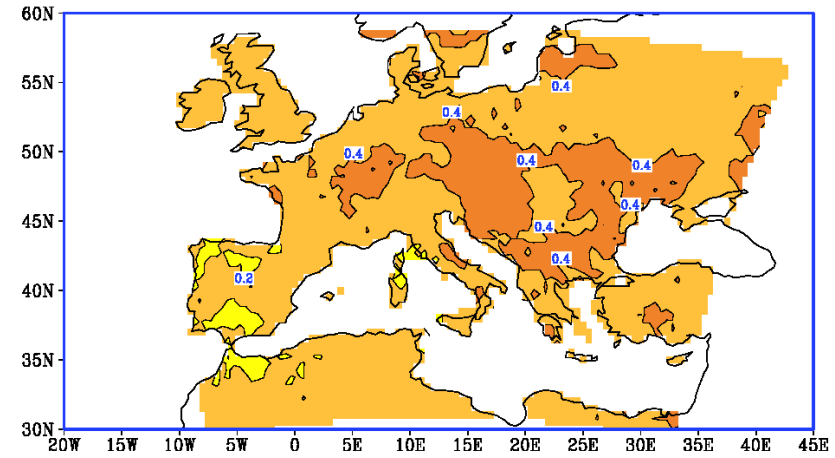
T2m anom>+0; JFM Threat score; ECMWF (CRU)  
cont=0.001 0.2 0.4 0.6 0.8 0.999



T2m anom>+0; JAS Threat score; RegCM (CRU)  
cont=0.001 0.2 0.4 0.6 0.8 0.999

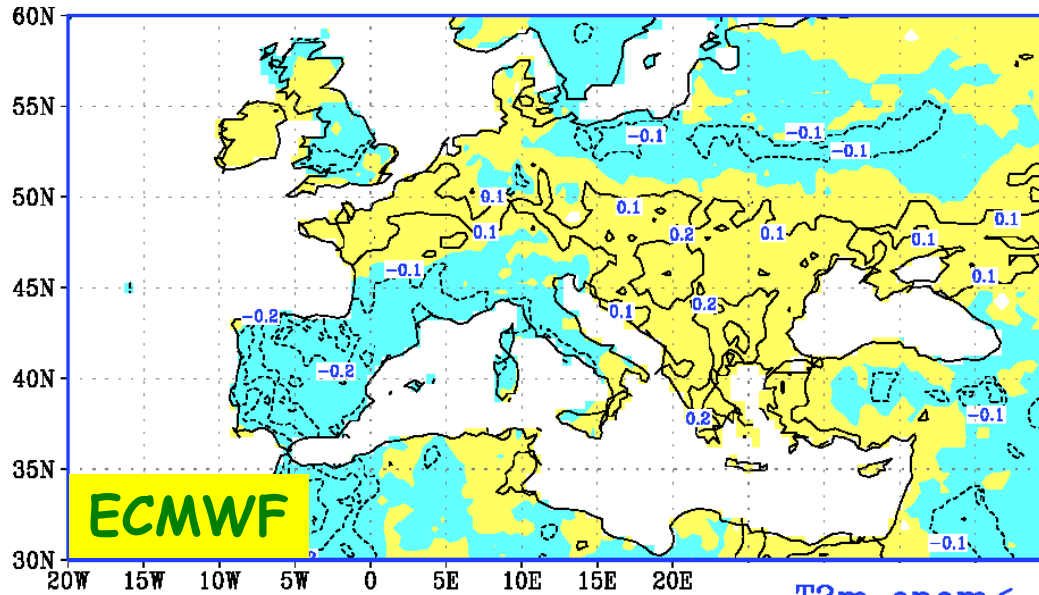


T2m anom>+0; JFM Threat score; RegCM (CRU)  
cont=0.001 0.2 0.4 0.6 0.8 0.999





T2m anom < -0; JFM Kuipers score; ECMWF (CRU)  
cont=0.1



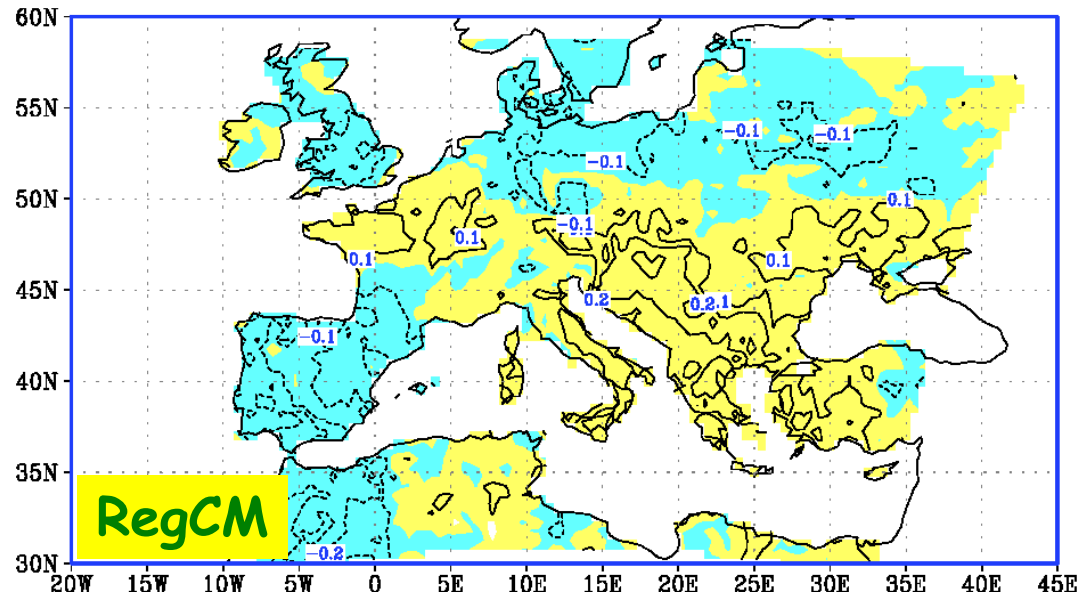
T2m anom < -0°C  
JFM

Kuipers score

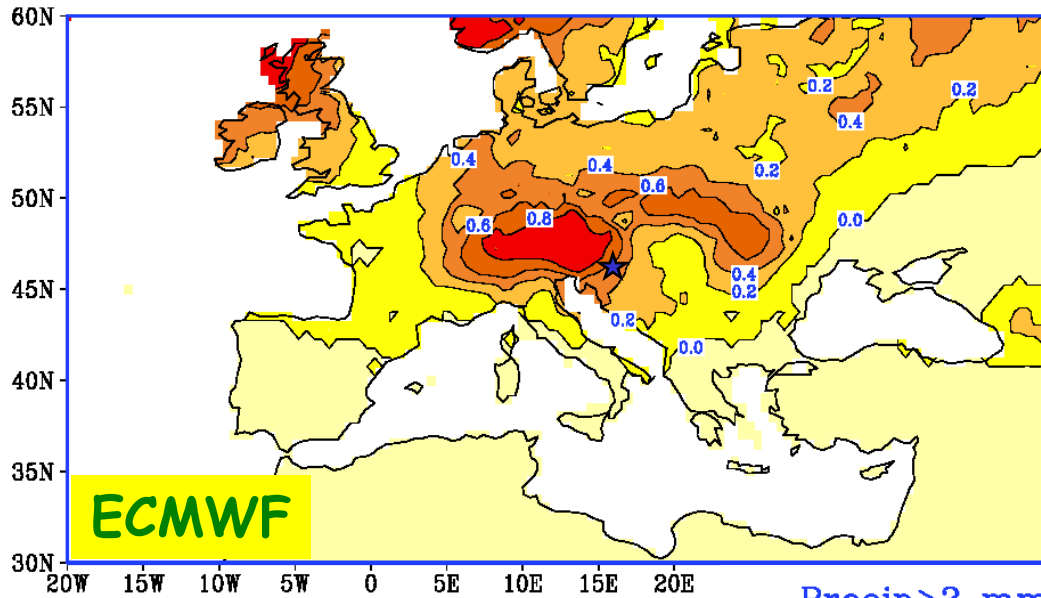
$$KSS = \frac{ad-bc}{(a+c)(b+d)}$$

		Observed	
		Y	N
Fcst	Y	a	b
	N	c	d

T2m anom < -0; JFM Kuipers score; RegCM (CRU)  
cont=0.1



Precip>2 mm/day; JAS Threat score; ECMWF  
 cont=0.001 0.2 0.4 0.6 0.8 0.999



Precip>2.0 mm/day  
 JAS

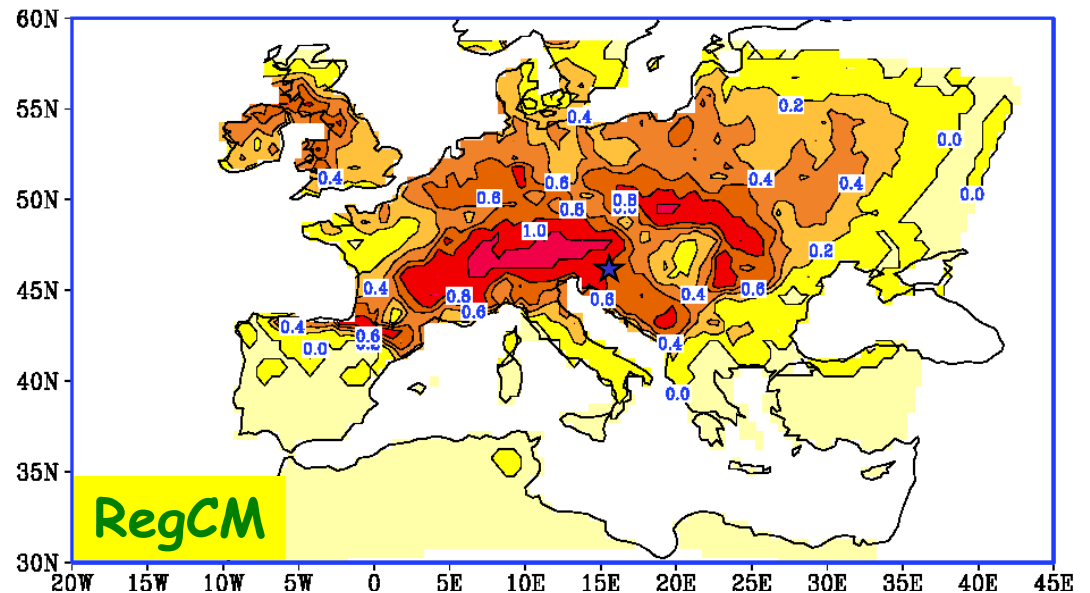
Threat score

Average for Zagreb  
 270 mm

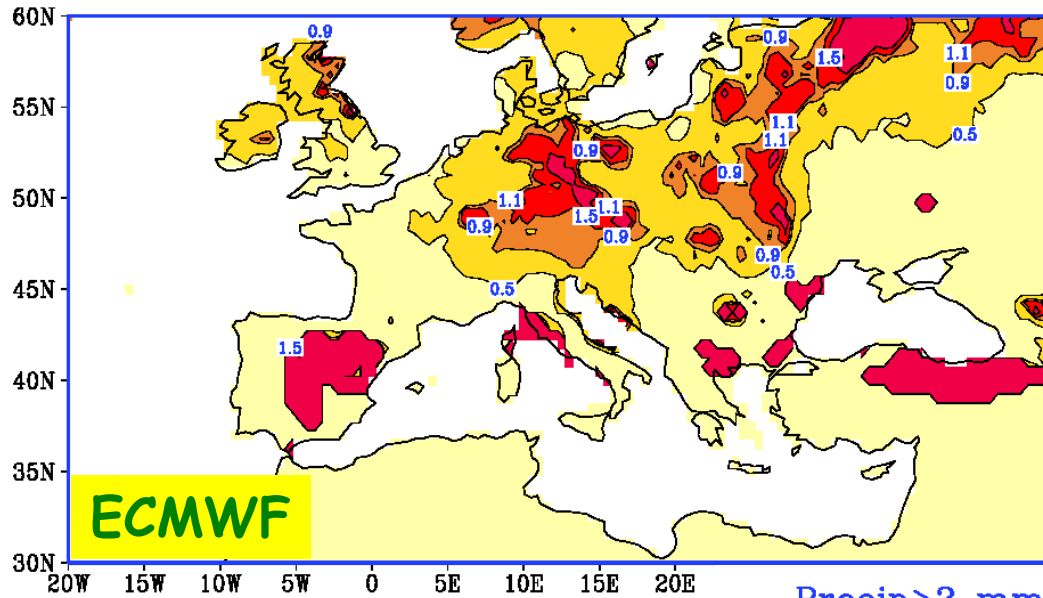
$$TS = \frac{a}{a+b+c}$$

		Observed	
		Y	N
Fcst	Y	a	b
	N	c	d

Precip>2 mm/day; JAS Threat score; RegCM  
 cont=0.001 0.2 0.4 0.6 0.8 0.999



Precip > 2 mm/day; JAS Bias; ECMWF  
cont=0.5 0.9 1.1 1.5



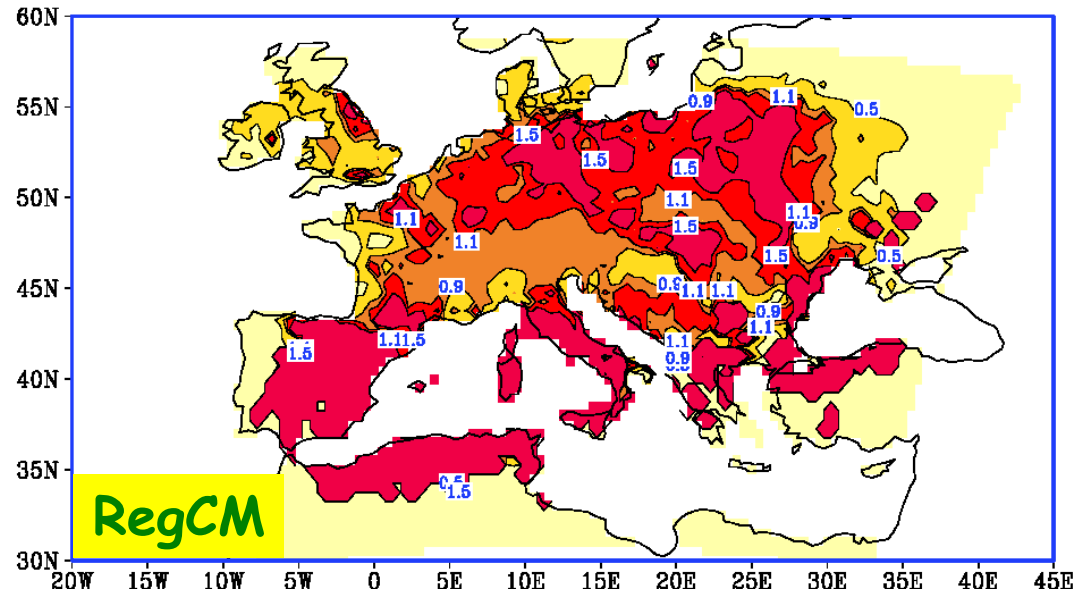
Precip > 2.0 mm/day  
JAS

Bias

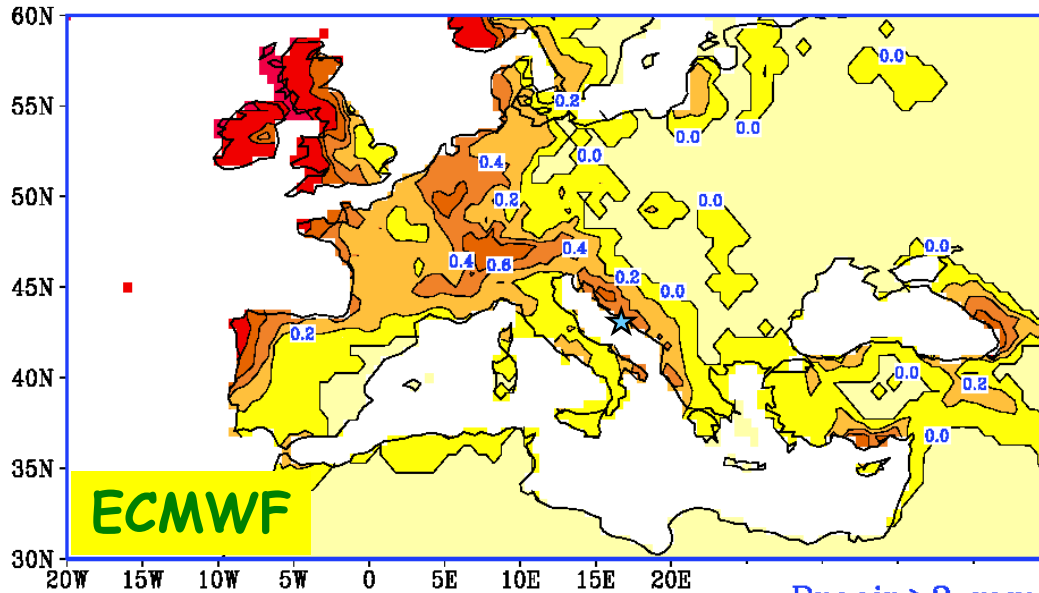
$$B = \frac{a+b}{a+c}$$

		Observed	
		Y	N
Fcst	Y	a	b
	N	c	d

Precip > 2 mm/day; JAS Bias; RegCM  
cont=0.5 0.9 1.1 1.5



Precip>2 mm/day; JFM Threat score; ECMWF  
 cont=0.001 0.2 0.4 0.6 0.8 0.999



Precip>2.0 mm/day  
 JFM

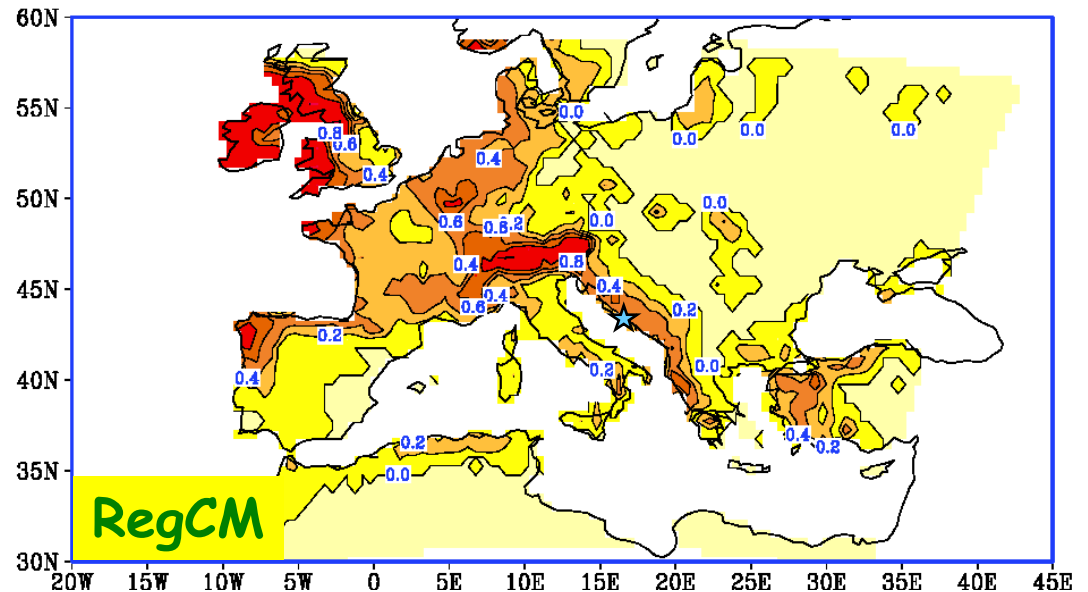
Threat score

Average for Split  
 156 mm

$$TS = \frac{a}{a+b+c}$$

		Observed	
		Y	N
Fcst	Y	a	b
	N	c	d

Precip>2 mm/day; JFM Threat score; RegCM  
 cont=0.001 0.2 0.4 0.6 0.8 0.999



## Some thoughts on verification statistics:

- \* No clear overall winner, but RM tends to be better for higher thresholds and over mountains  
(results may improve in favour of RM with a higher resolution)
- \* Need to know better systematic biases of RM  
("climate" of 1990's is biased)
- \* How to best verify results of downscaling ?

... and some thoughts on dynamical downscaling:

- \* Probably not worth the trouble for upper-air fields (?)
- \* Improves the structure of surface fields
- \* If GCM forecast is good, a significant benefit of downscaling in orography-related fields  
(need for ever improved orography)
- \* Won't improve bad global forecast
- \* It is as good as RM is good