

# Precipitation verification

Contributions from: CMC, CPTEC, DWD, ECMWF, JMA, MF, NCEP, NRL, RHMC, UKMO



WGNE - 31  
Council for Scientific and Industrial Research  
South Africa  
26 - 29 April 2016



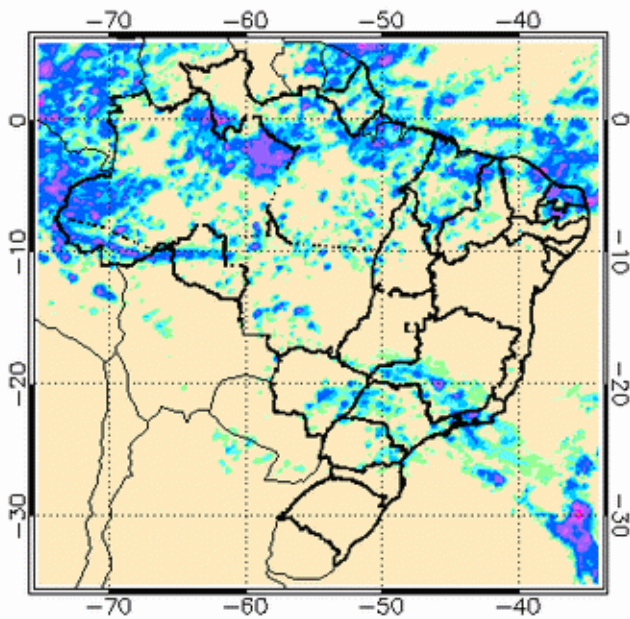
# Operational verification of quantitative precipitation forecast at INPE/CPTEC

April, 2016

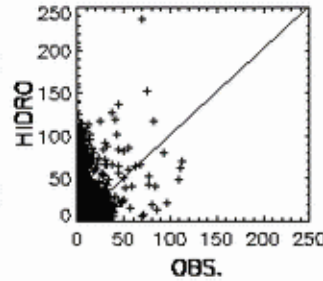
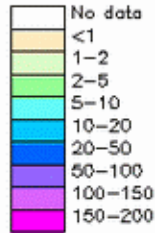
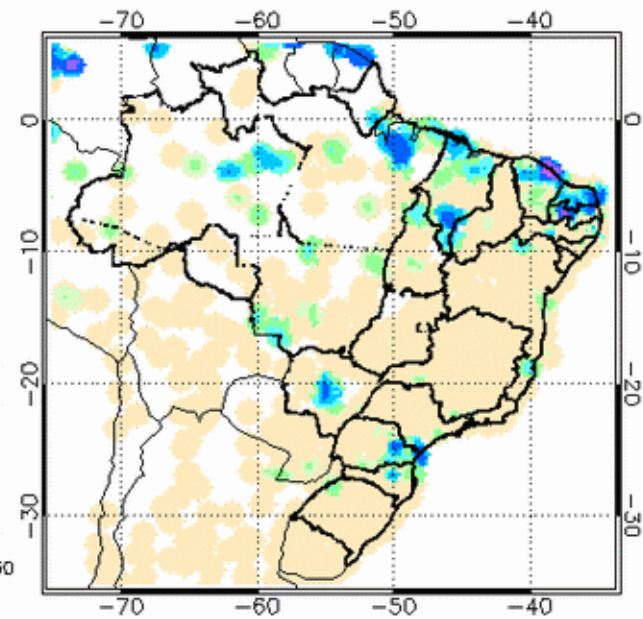
Thanks for José R. Rozante and Daniel A. Vila,

# Satellite precipitation validation

HIDRO estimates for 20160331

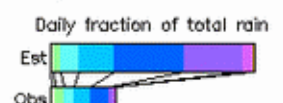
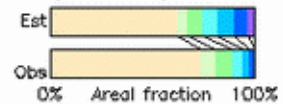


Daily gauge analysis (land only) for 20160331



Rain gauges in South America  
White areas: no rain gauges stations

Daily fraction by occurrence



Rainfall accumulation by amount



HIDRO

	<1	≥1
<1	4554	1787
≥1	800	1402

Verification statistics for 20160331 n=8543 Verif. grid=0.25° Units=mm/d

	Observed	HIDRO	
# gridpoints raining	2202	3189	Mean abs error = 6.1
Average rain	2.1	6.4	RMS error = 14.9
Conditional rain	8.0	17.2	Correlation coeff = 0.352
Rain volume (mm*km <sup>2</sup> *10 <sup>9</sup> )	12.5	39.0	Frequency bias = 1.448
Maximum rain	111.9	237.2	Probability of detection = 0.637
			False alarm ratio = 0.560
			Hansen & Kuipers score = 0.355
			Equitable threat score = 0.183

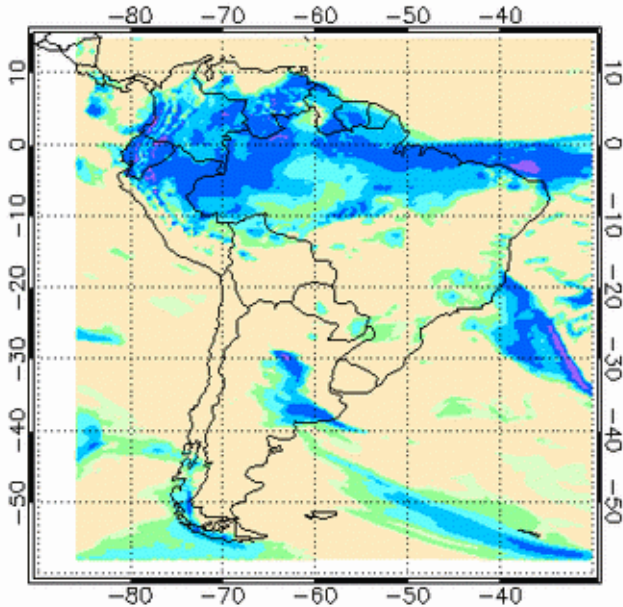
Statistical scores

Hydroestimator – comparison with rain gauges

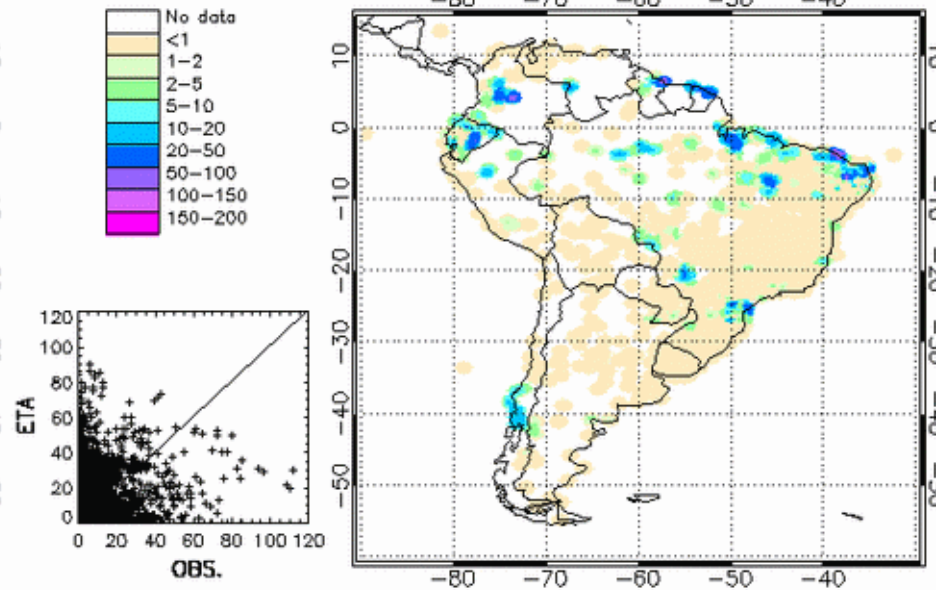
# Model precipitation validation

Regional  
model  
ETA

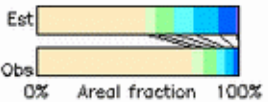
ETA estimates for 20160331



Daily gauge analysis (land only) for 20160331



Daily fraction by occurrence



Daily fraction of total rain



Rainfall accumulation by amount

	ETA		Observed
	<1	≥1	
<1	7986	4796	
≥1	1069	2723	

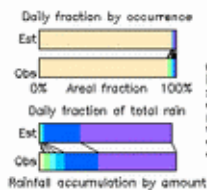
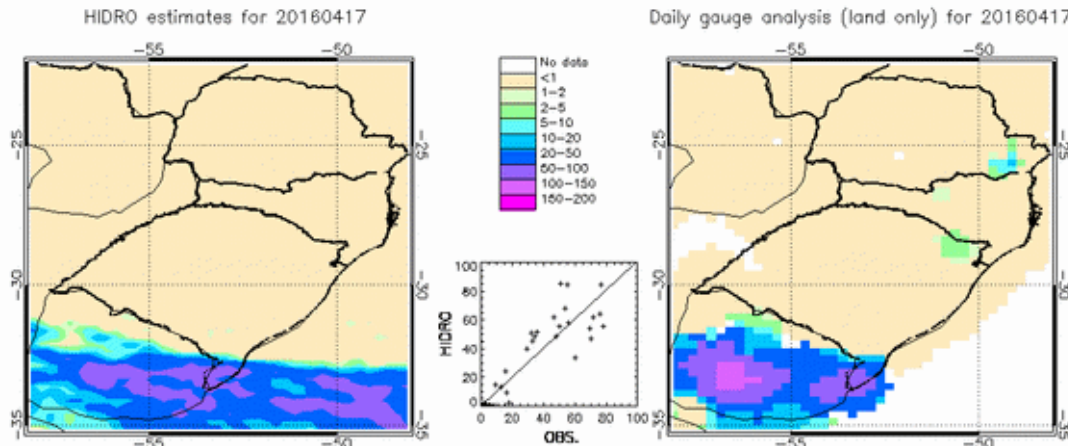
Verification statistics for 20160331 n=16574 Verif. grid=0.25° Units=mm/d

	Observed	ETA	
# gridpoints raining	3792	7519	Mean abs error = 5.8
Average rain	2.1	6.0	RMS error = 11.2
Conditional rain	9.0	13.2	Correlation coeff = 0.279
Rain volume (mm*km <sup>2</sup> *10 <sup>6</sup> )	24.3	70.5	Frequency bias = 1.983
Maximum rain	111.9	90.4	Probability of detection = 0.718
			False alarm ratio = 0.638
			Hanssen & Kuipers score = 0.343
			Equitable threat score = 0.146



# The same for different areas

## Southern Brazil



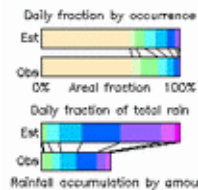
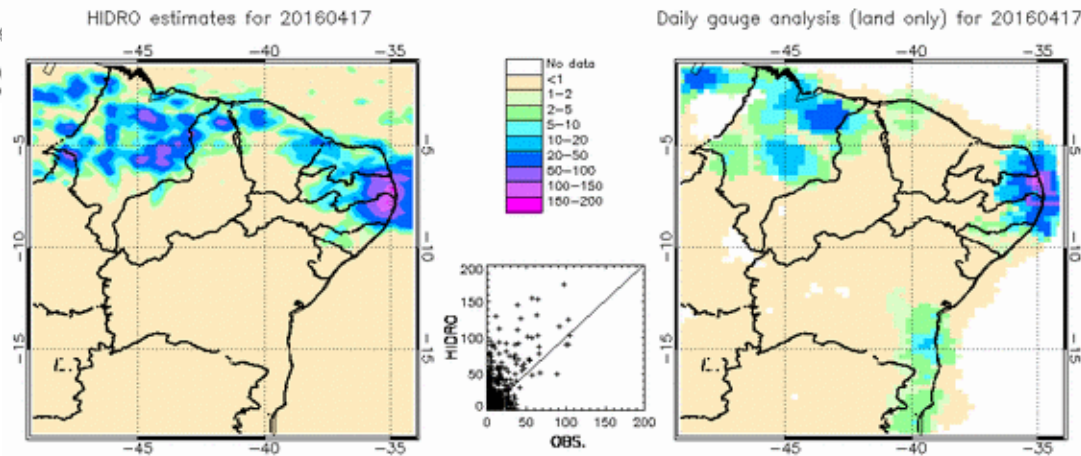
		HIDRO	
		<1	≥1
Obs	<1	879	5
	≥1	38	25

Verification statistics for 20160417 n=947 Verif. grid=0.25° Units=mm/d

	Observed	HIDRO
# gridpoints raining	63	30
Average rain	1.3	1.3
Conditional rain	20.2	40.9
Rain volume (mm*km <sup>2</sup> *10 <sup>6</sup> )	0.9	0.9
Maximum rain	78.1	85.7

Mean abs error = 0.6  
 RMS error = 3.1  
 Correlation coeff = 0.932  
 Frequency bias = 0.476  
 Probability of detection = 0.31  
 False alarm ratio = 0.167  
 Hanssen & Kuipers score = 0  
 Equitable threat score = 0.349

## Northeast Brazil



		HIDRO	
		<1	≥1
Obs	<1	1123	253
	≥1	219	508

Verification statistics for 20160417 n=2103 Verif. grid=0.25° Units=mm/d

	Observed	HIDRO
# gridpoints raining	727	761
Average rain	3.4	6.7
Conditional rain	9.7	18.5
Rain volume (mm*km <sup>2</sup> *10 <sup>6</sup> )	5.0	10.0
Maximum rain	104.3	174.7

Mean abs error = 5.2  
 RMS error = 13.1  
 Correlation coeff = 0.688  
 Frequency bias = 1.047  
 Probability of detection = 0.699  
 False alarm ratio = 0.332  
 Hanssen & Kuipers score = 0.515  
 Equitable threat score = 0.342

# Model verification

<http://avaliacaodemodelos.cptec.inpe.br/>

Monthly categorical  
verification

FAR for South America

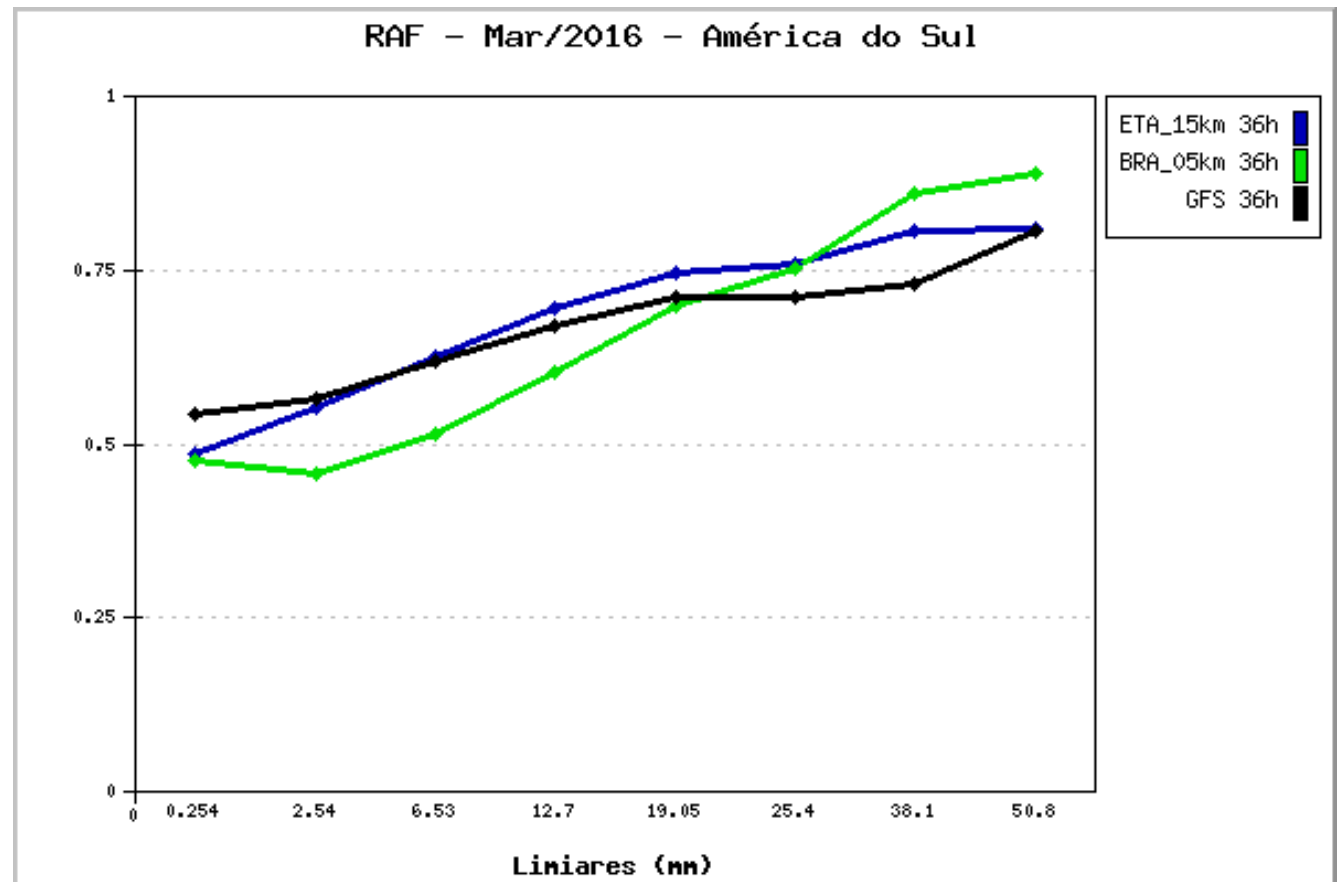
Models evaluated:

BRAMS

ETA 

GFS 





# Model verification

<http://avaliacaodemodelos.cptec.inpe.br/>

Monthly categorical  
verification

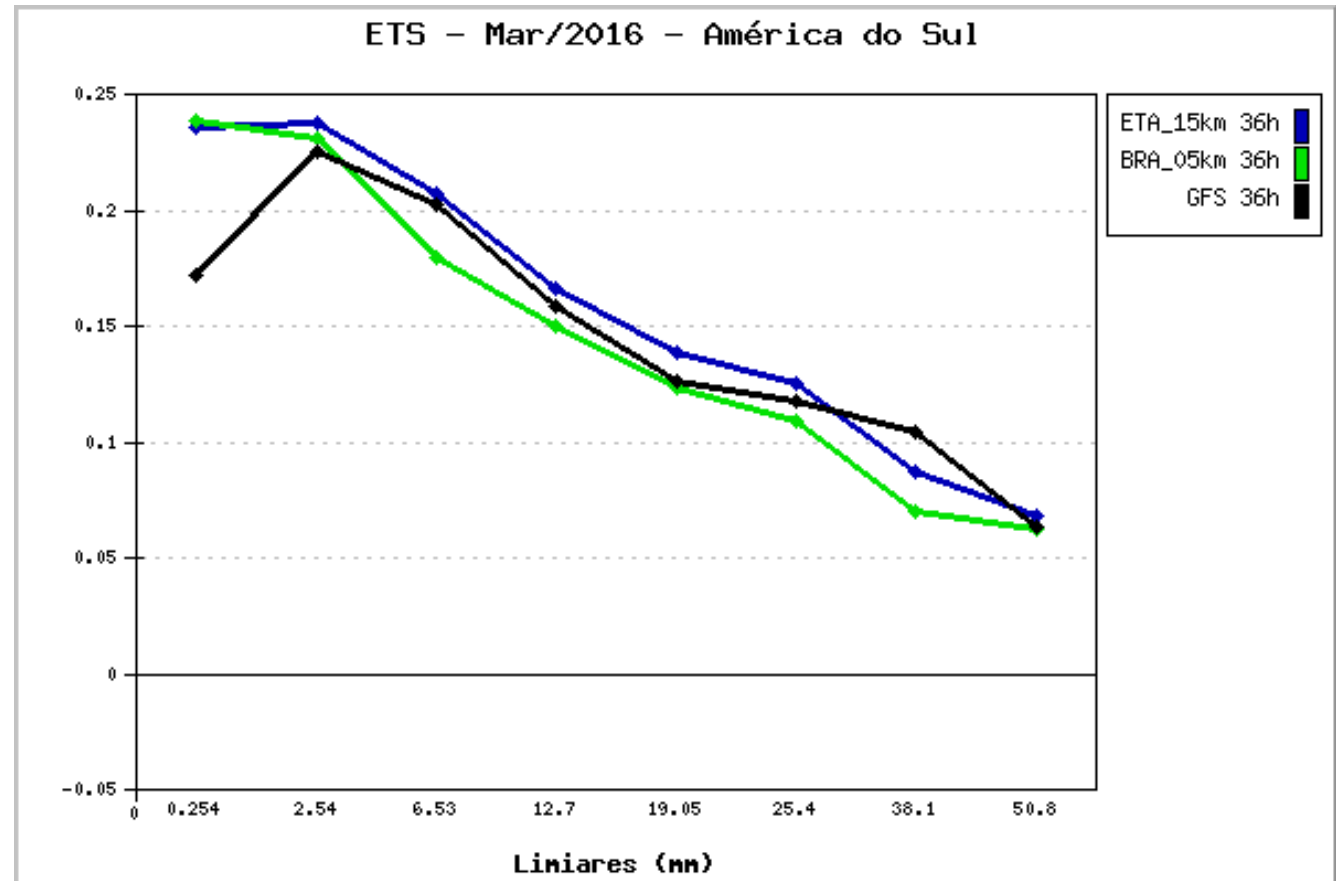
ETS for South America

Models evaluated:

BRAMS

ETA

GFS



# DWD

22 April 2016

Felix Fundel, Ulrich Pflüger (DWD)



# Verification Settings

## Participating

- Germany: DWD GME (until Dec. 2014) and ICON (from Jan. 2015)
- ECMWF: IFS
- France: Meteo-France Arpege
- Canada: CMC GDPS
- U.K.: UKMO Unified Model (no data for DJF 2014)
- ~~USA: NCEP GFS (data problems)~~

## Observations

- Calibrated radar composite over Germany
- 24h sums (06-06 UTC)

## Preprocessing

- Re-gridding to  $0.025^{\circ} \times 0.025^{\circ}$  (both models and observations)

## Verification

- Scales 1x1, 3x3, 5x5, 9x9 grid points
- Thresholds 0.1, 1, 2, 5, 10, 20, 50 mm/24h
- Lead-Times: 30, 54 and 78h of 00 UTC runs
- Seasonal verification DJF 2014 – DJF 2015

Seasonal avg. of  
daily  
precipitation

LT 06-30h

**!DATA PROBLEMS!**

**RADAR**

**GERMANY**

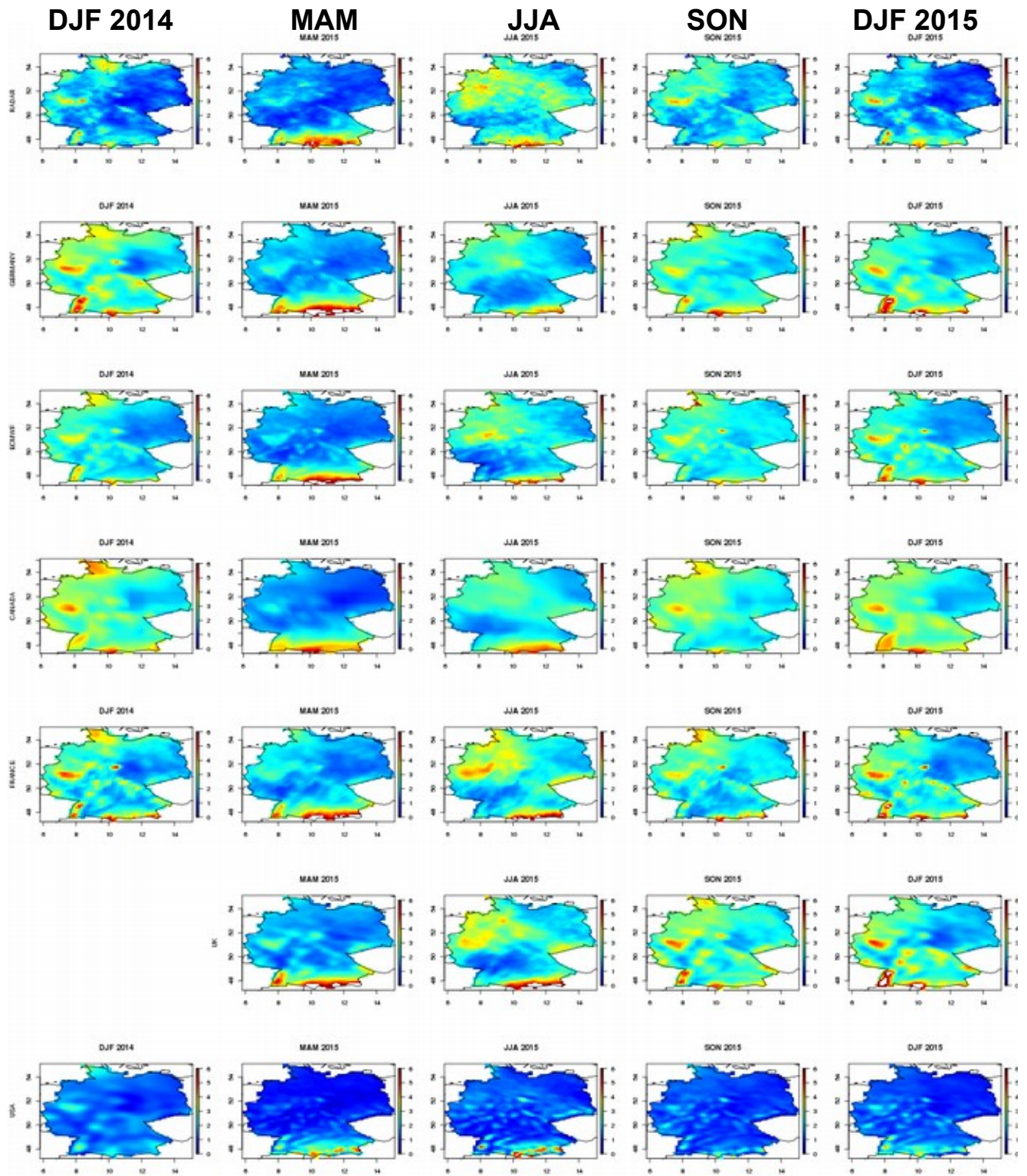
**ECMWF**

**CANADA**

**FRANCE**

**UK**

**USA**



# Fractions Skill Score

JJA 2015

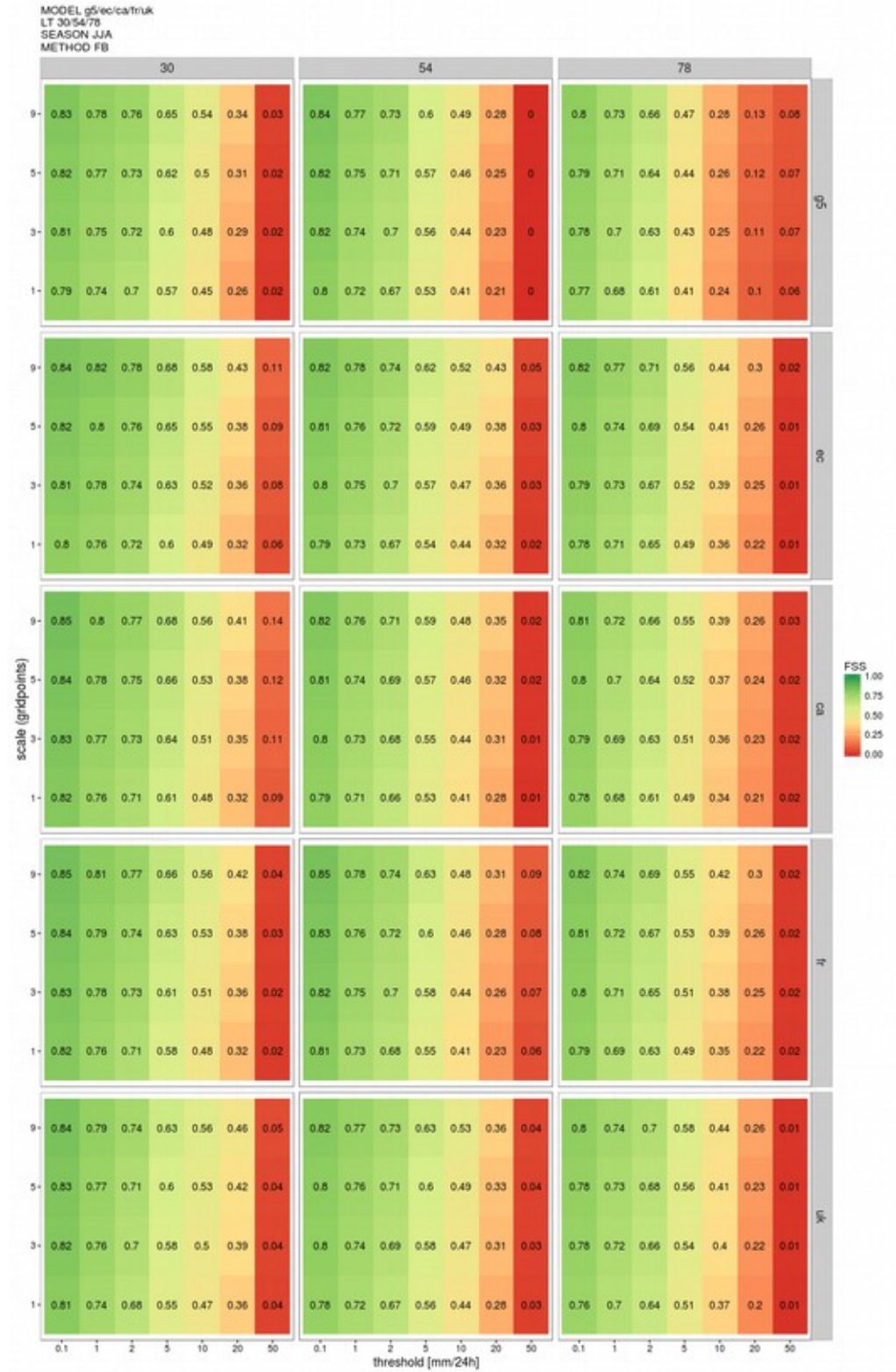
GERMANY

ECMWF

CANADA

FRANCE

U.K.





# Fractions Skill Score

DJF 2015

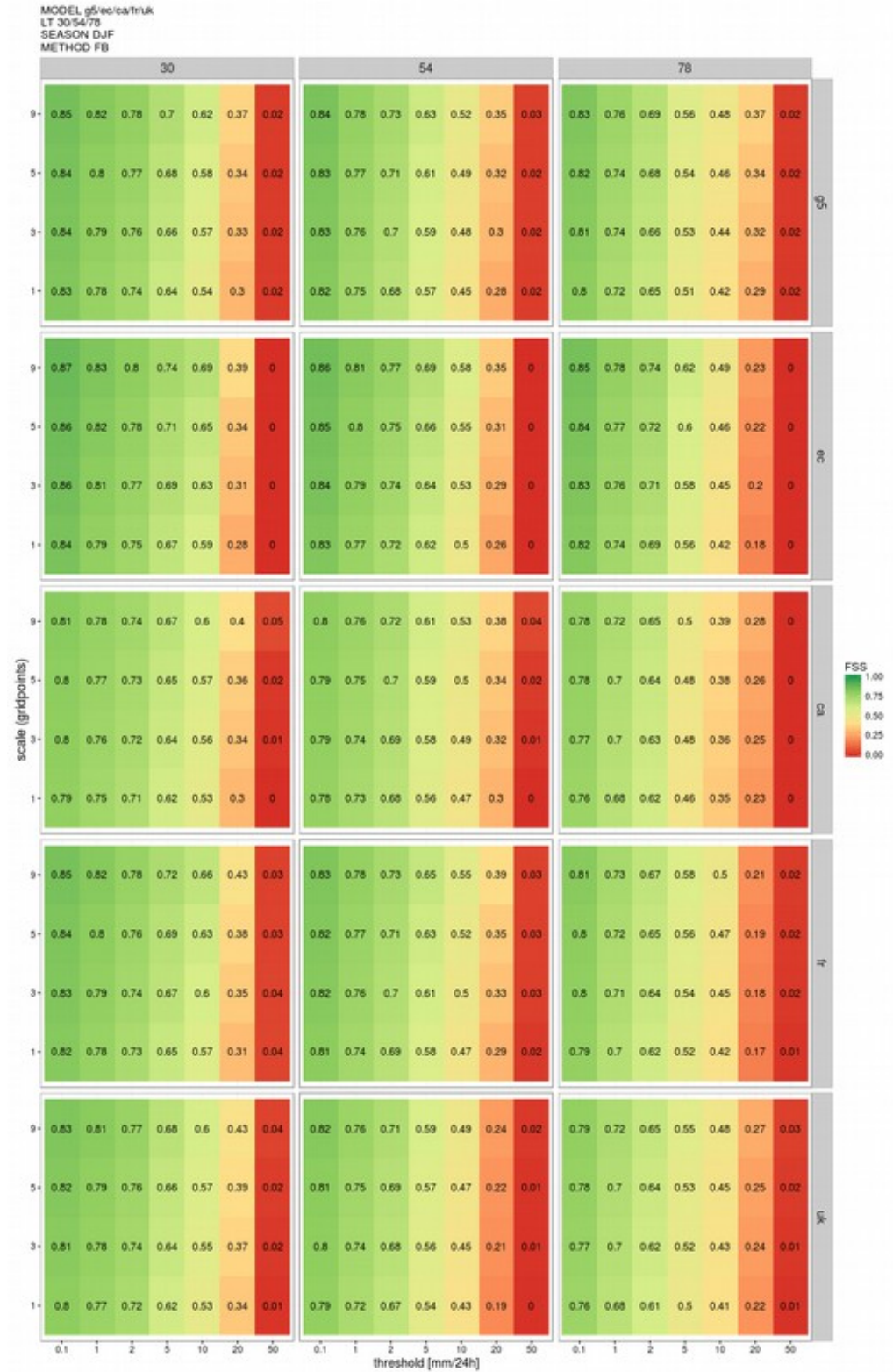
GERMANY

ECMWF

CANADA

FRANCE

U.K.





MODEL g5ec/ca/truk  
LT 30/54/78  
SEASON JJA  
METHOD PG

# Brier Skill Score (pragmatic)

## JJA 2015

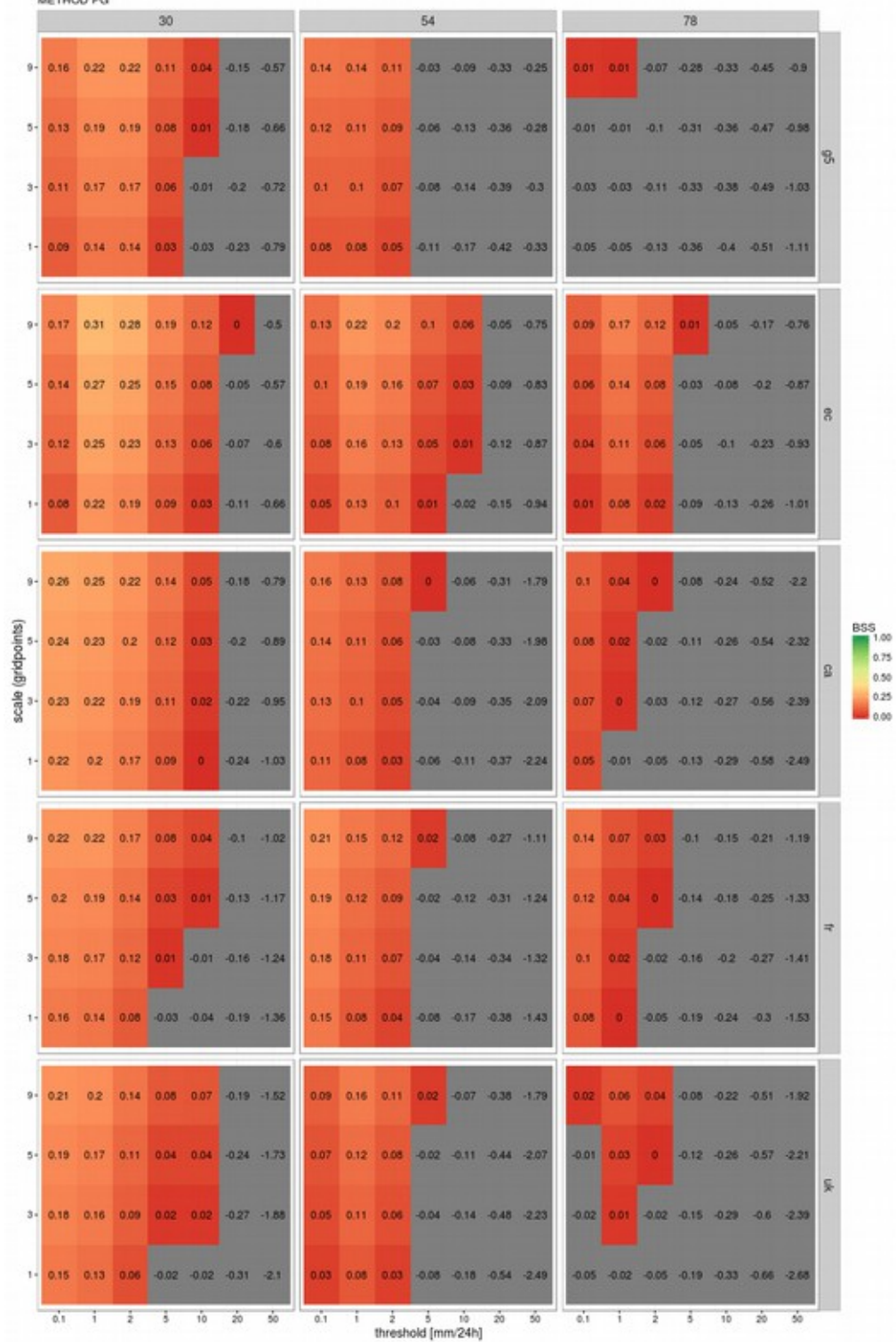
### GERMANY

### ECMWF

### CANADA

### FRANCE

### U.K.

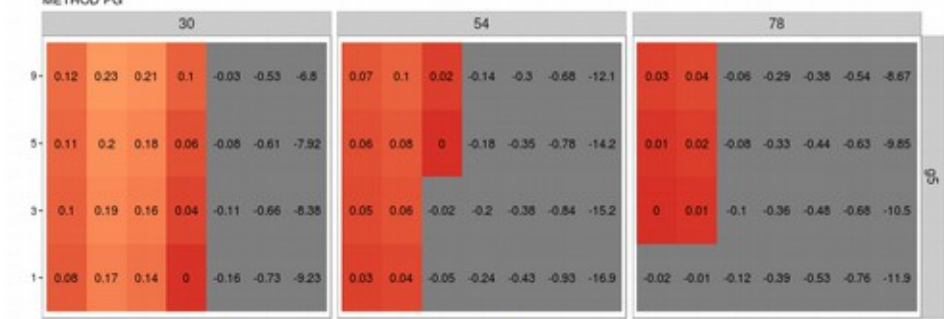


MODEL g5ec/ca/truk  
LT 30/54/78  
SEASON DJF  
METHOD PG

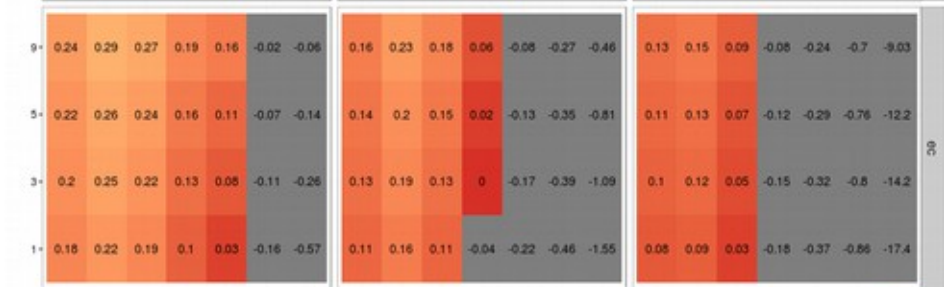
# Brier Skill Score (pragmatic)

## DJF 2015

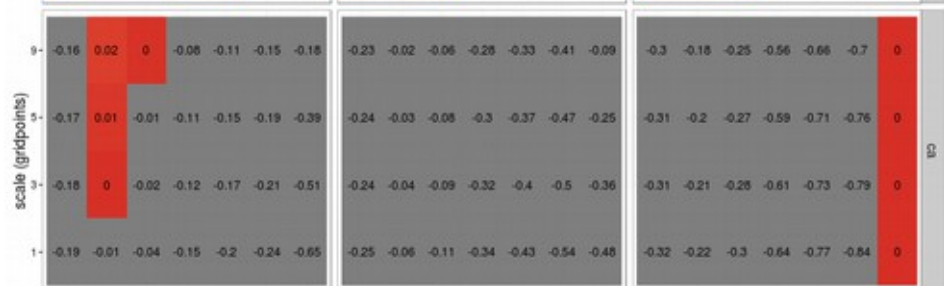
### GERMANY



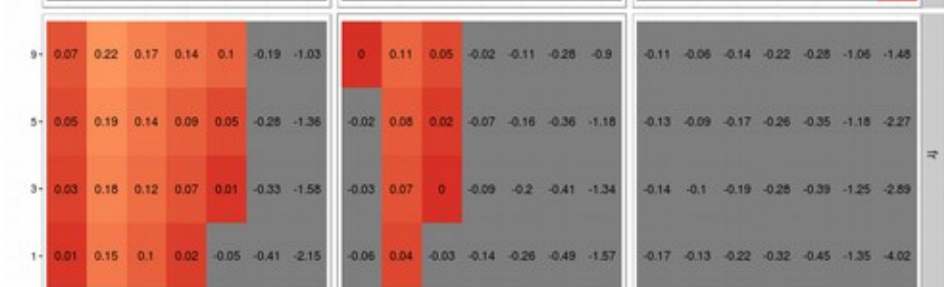
### ECMWF



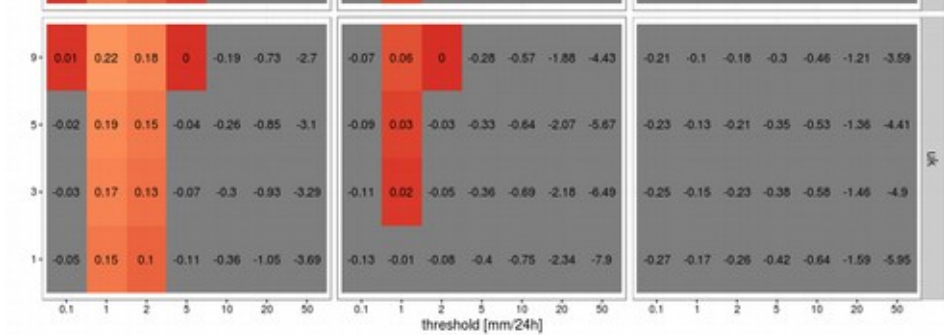
### CANADA



### FRANCE

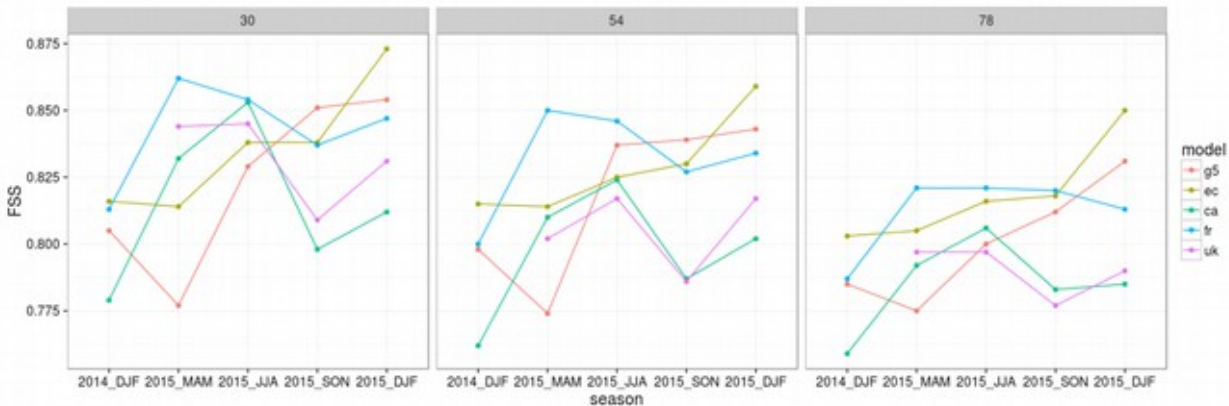


### U.K.

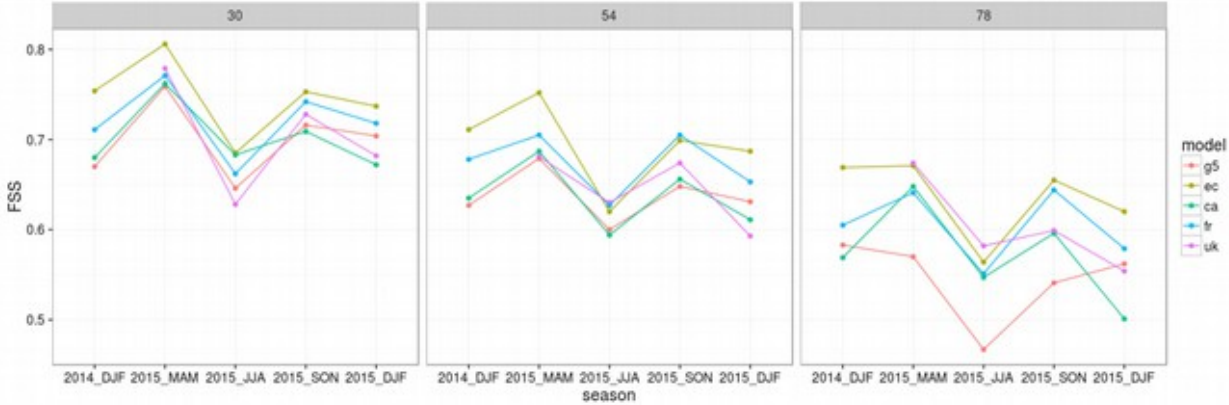


# Fractions Skill Score Scale: 9x9 GP

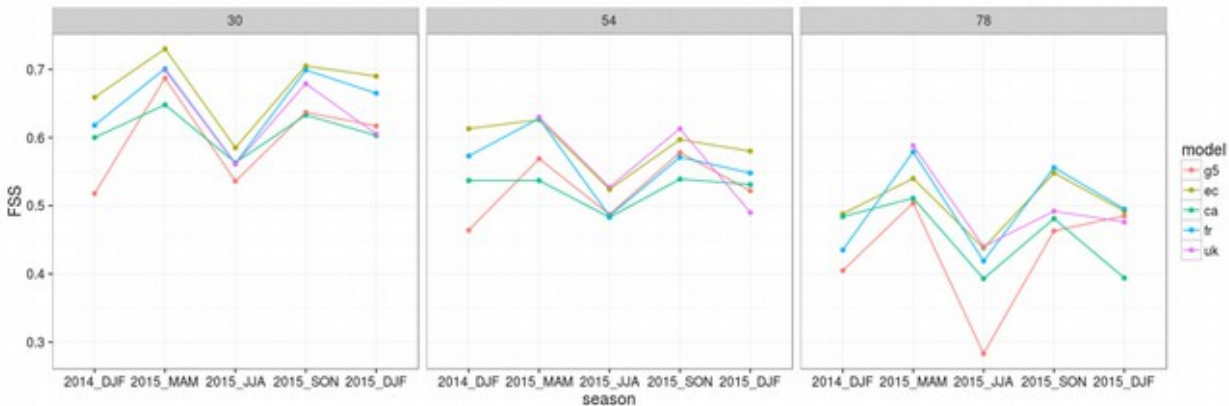
0.1 mm/24h



5 mm/24h

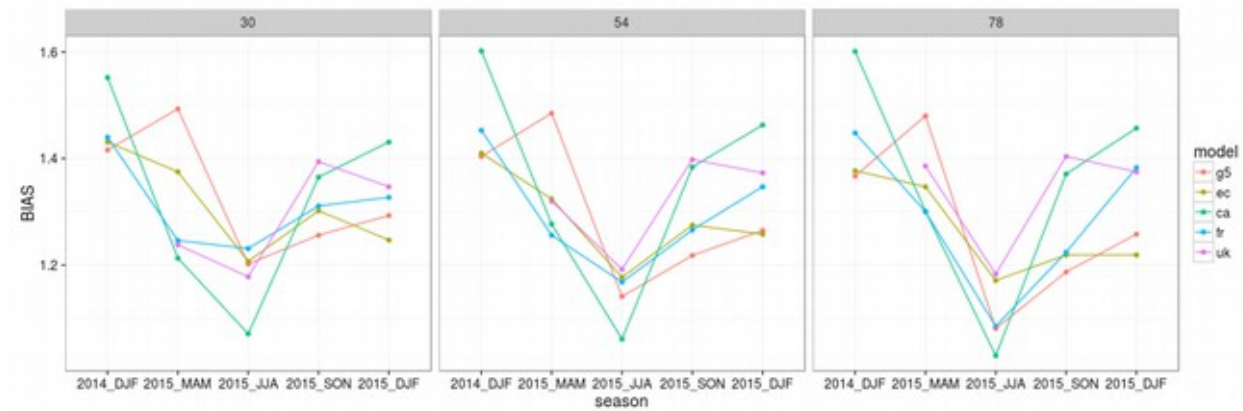


10 mm/24h

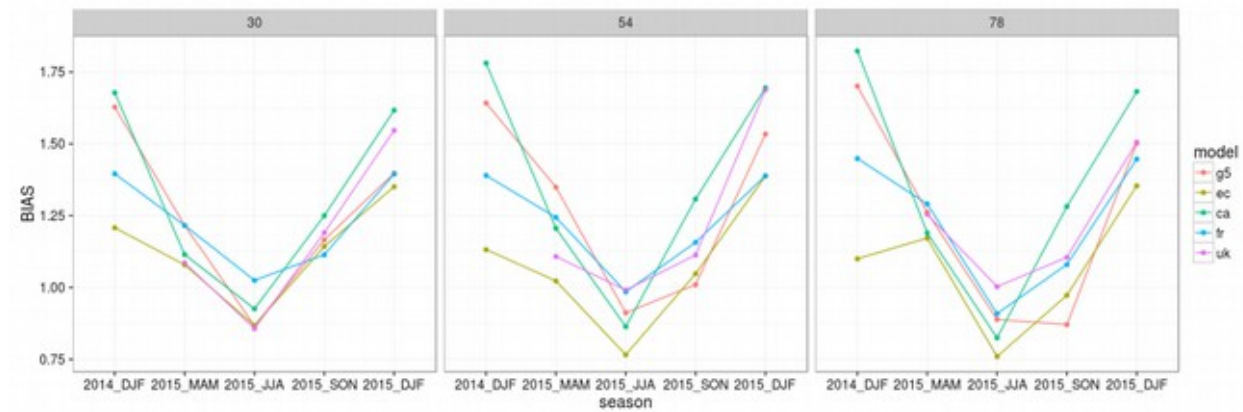


# Freq. Bias (upscaling) Scale: 9x9 GP

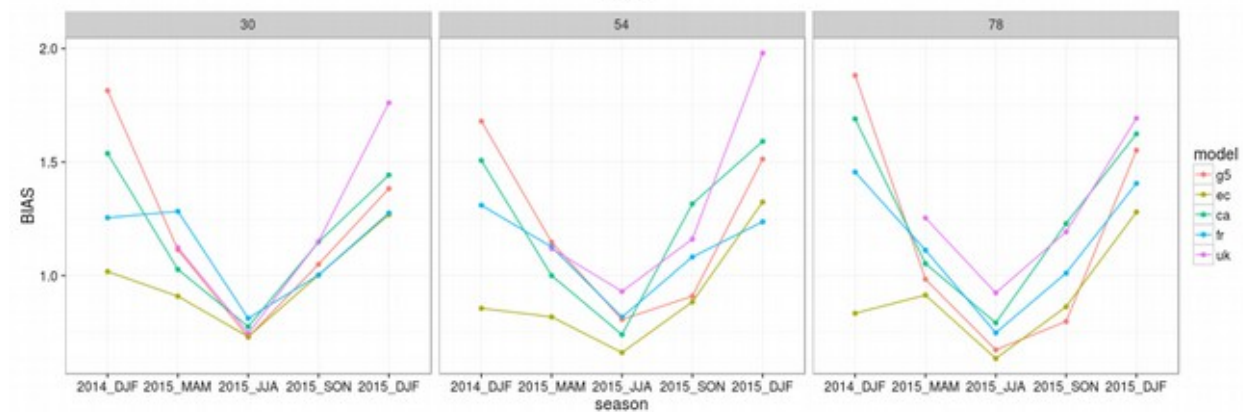
0.1 mm/24h



5 mm/24h



10 mm/24h





# ECMWF

Tomas Haiden

# Verification using SYNOP

## Characteristics

- 24-h precipitation
  - Forecast days 1 to 10
  - Aggregation over large domains (extra-tropics, tropics, Europe)

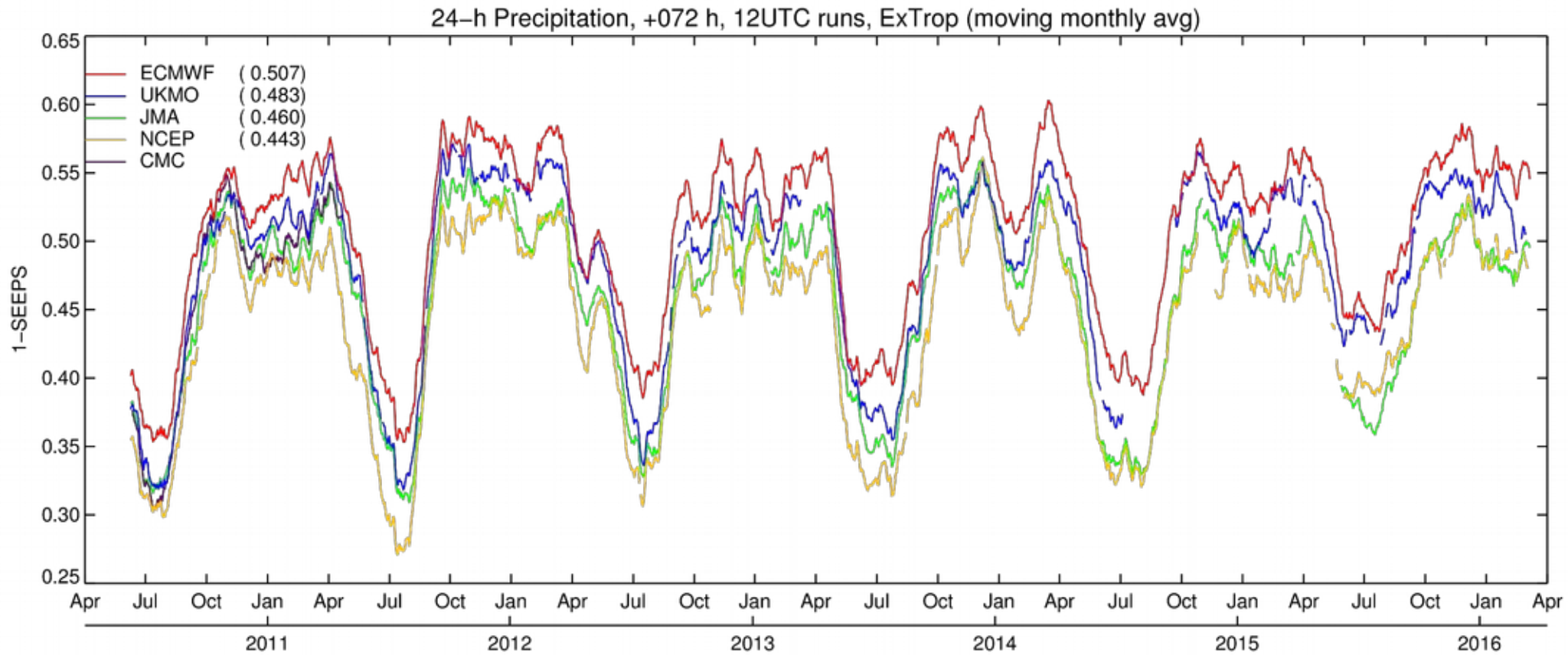
## Verification of Deterministic Forecasts

- Symmetric Equitable Error in Probability Space (SEEPS)
- Equitable Threat Score (ETS)
- Frequency bias (FB)
- Symmetric Extremal Dependence Index (SEDI)

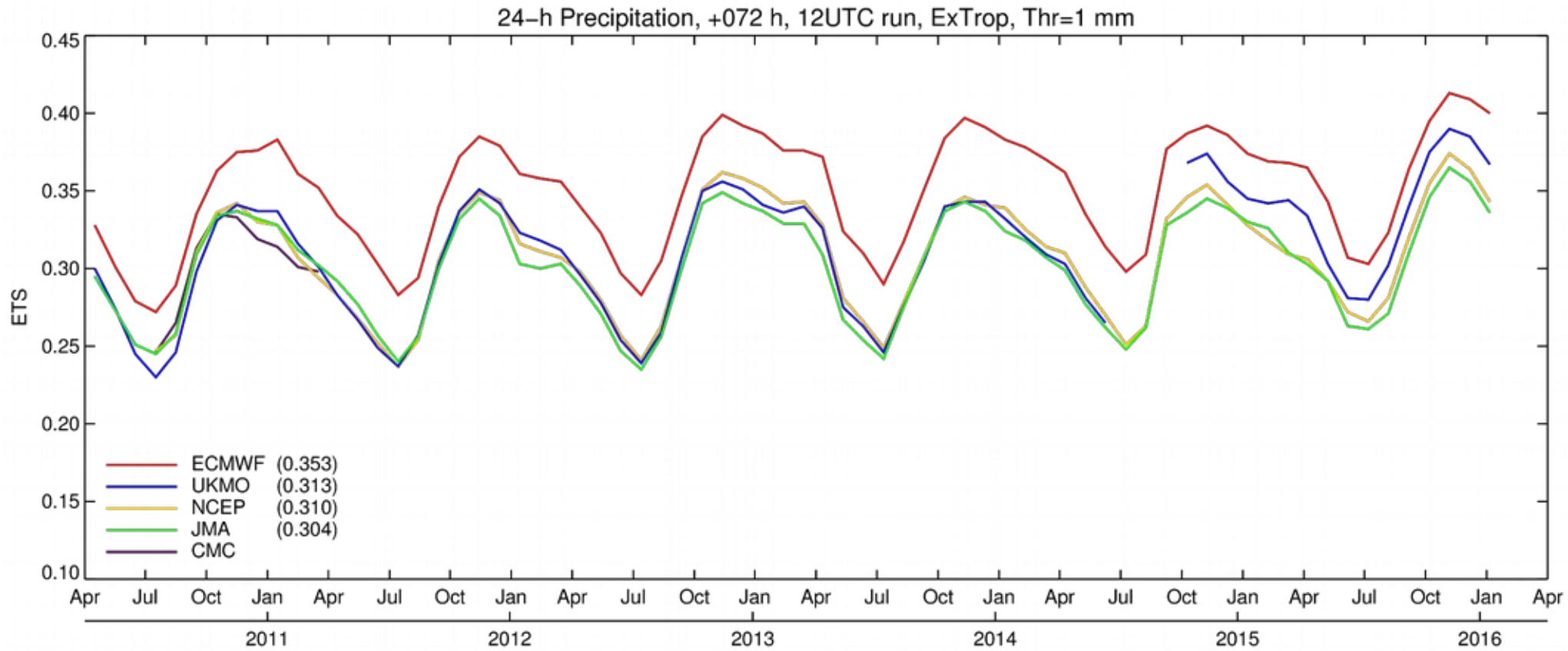
## Verification of Ensemble Forecasts

- Continuous Rank Probability Skill Score (CRPSS)
  - Brier Skill Score (BSS)

# Model intercomparison – deterministic forecast

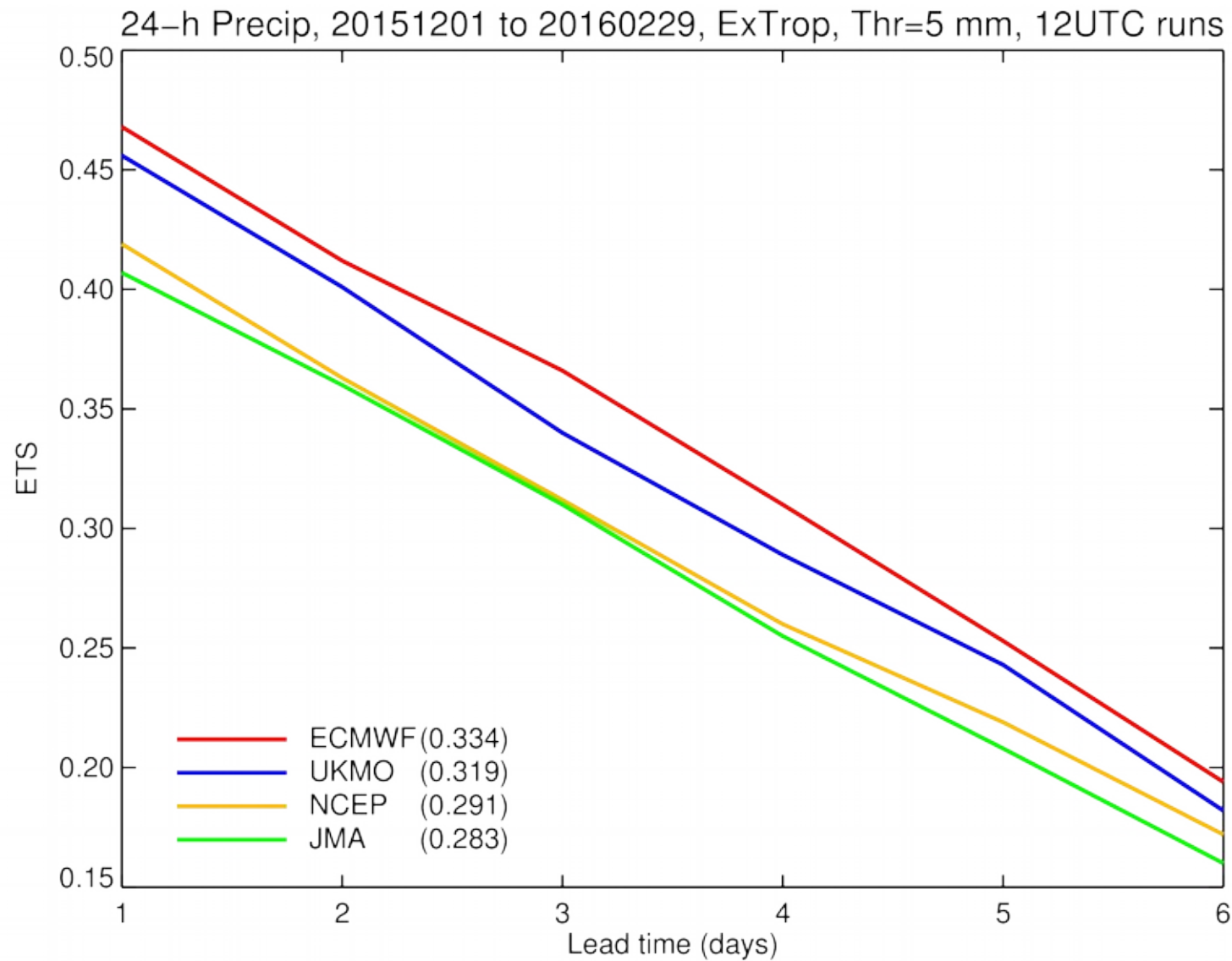


# Model intercomparison – deterministic forecast

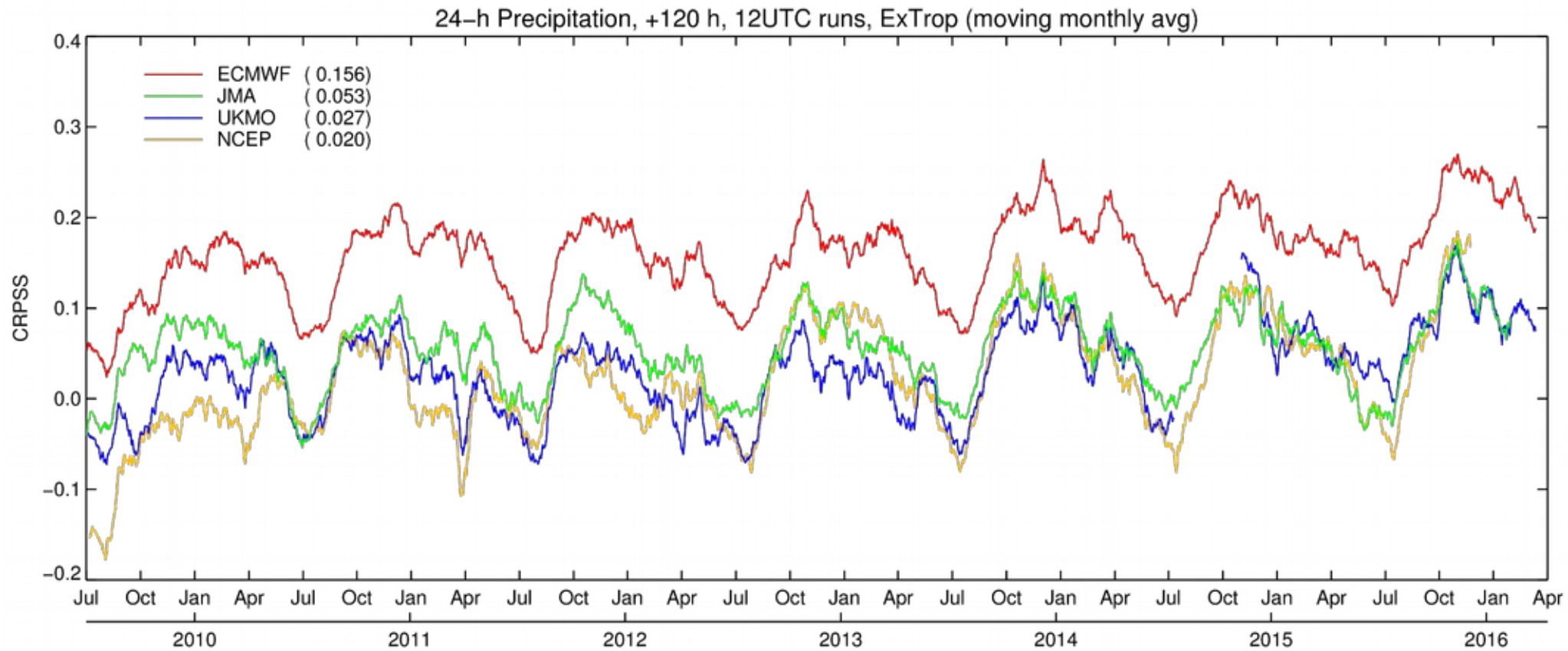




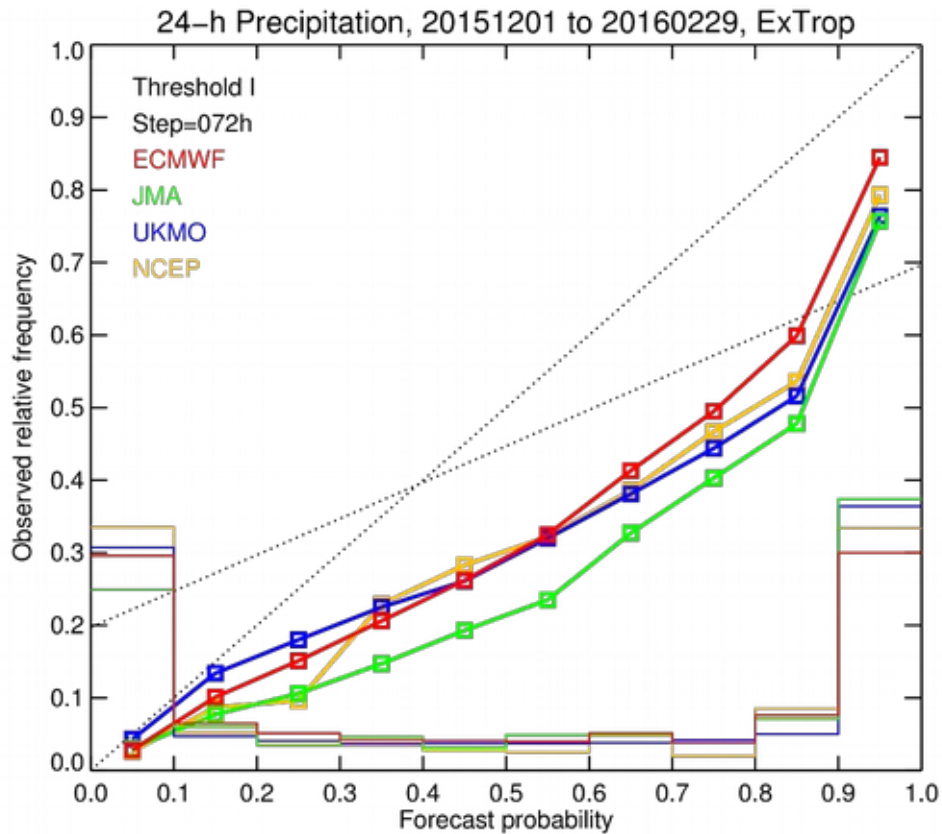
# Model intercomparison – deterministic forecast



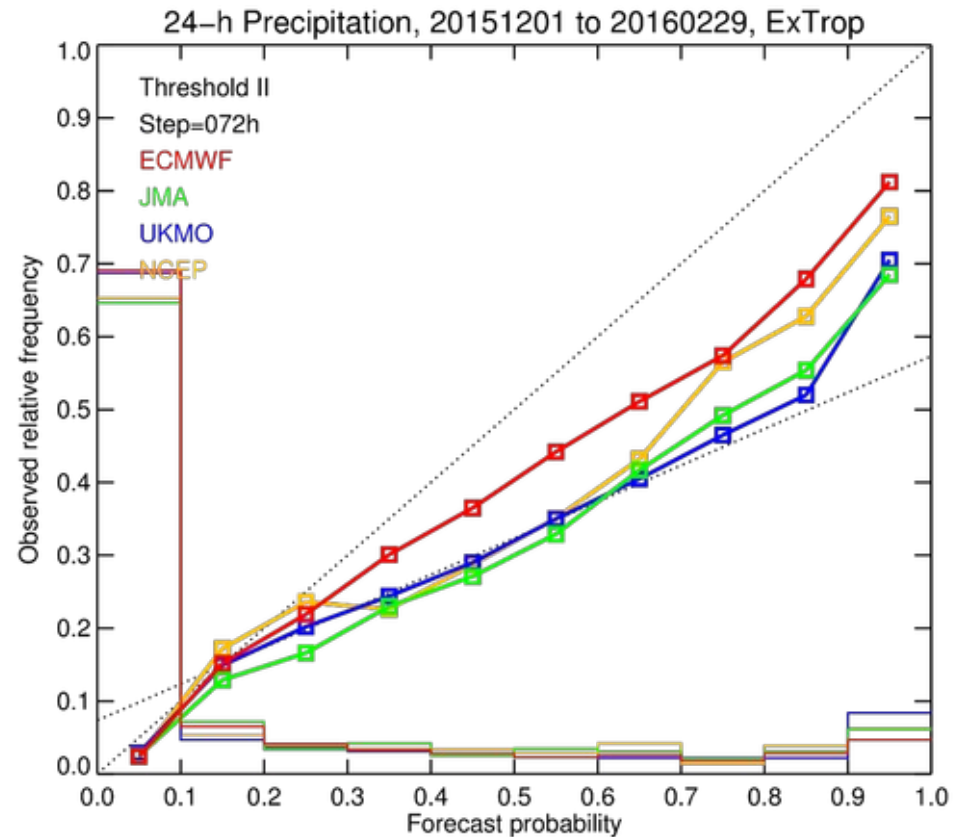
# Model intercomparison – ensemble forecast



# Model intercomparison – ensemble forecast



Light precipitation



Moderate-to-heavy  
precipitation

# Verification using additional datasets

## High Density observations (HDOBS)

- precipitation data from 15 ECMWF Member States
  - mix of hourly, 6-hourly, and daily reports
  - up to 16 months of data
  - HDOBS/SYNOP ratio of station numbers on average 3/1

## Verification of Deterministic Forecasts

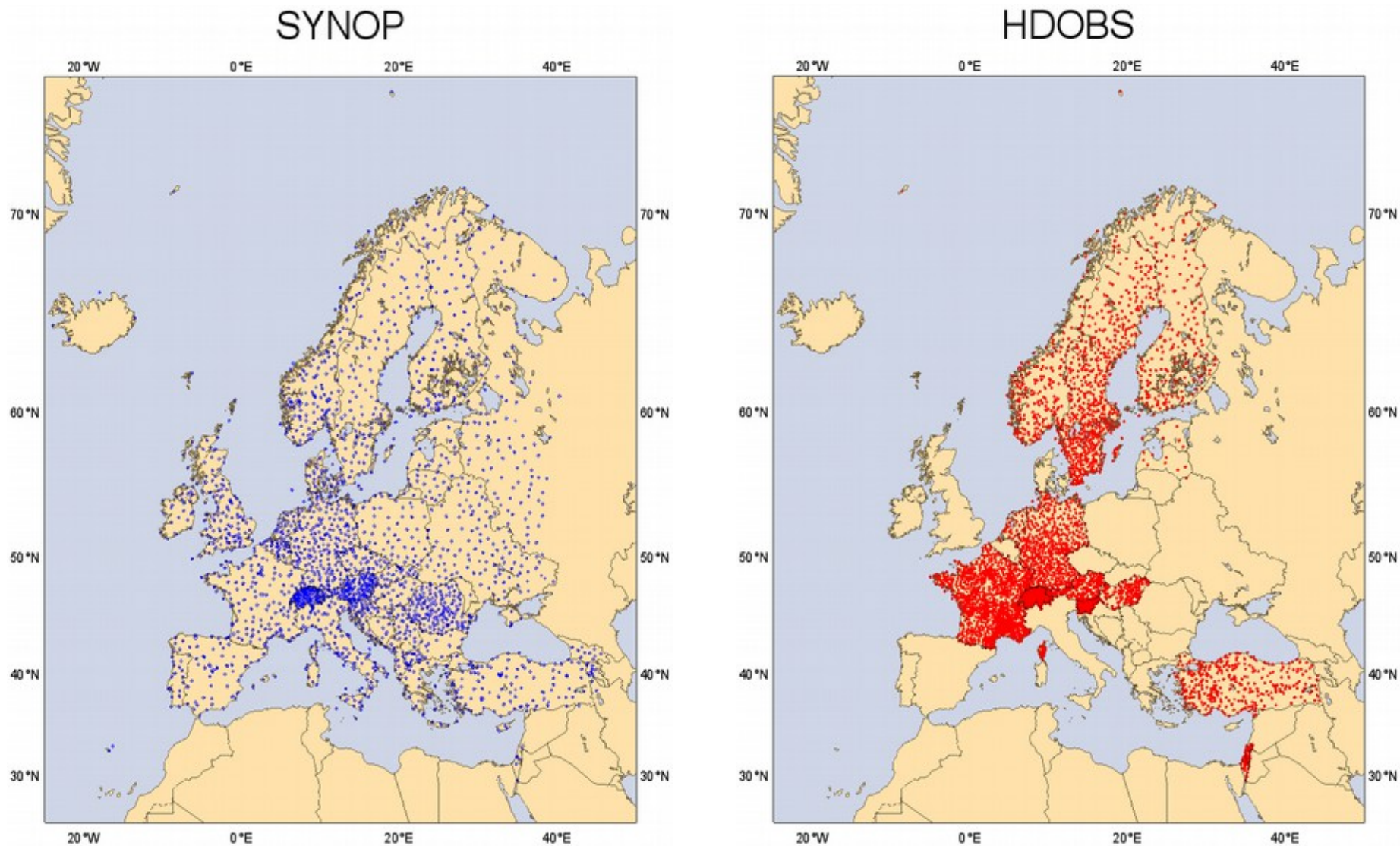
- Equitable Threat Score (ETS)
- Symmetric Extremal Dependence Index (SEDI)

## Verification of Ensemble Forecasts

- Continuous Rank Probability Score (CRPS)
- Brier Score (BS)



# Surface Observations – Daily Precip to 06Z

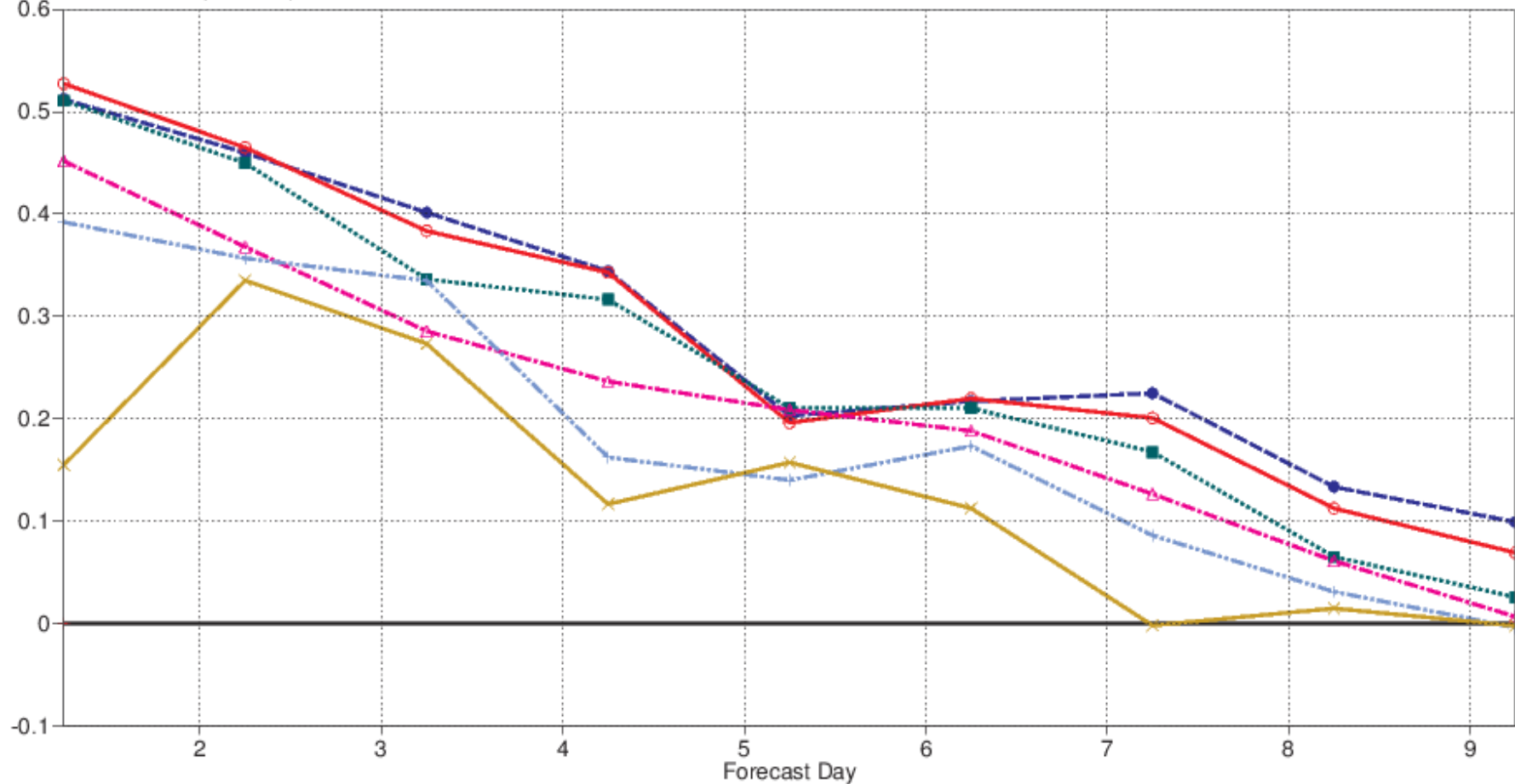


# HDOBS ETS for Austria in DJF 2015/16

total precipitation  
Equitable Threat Score  
Austria

Date: 201512 to 201602

HDOBS\_Austria od oper 0001 | Mean method: fair



# HDOBS 41r1-v-41r2 for France, DJF 2015/16

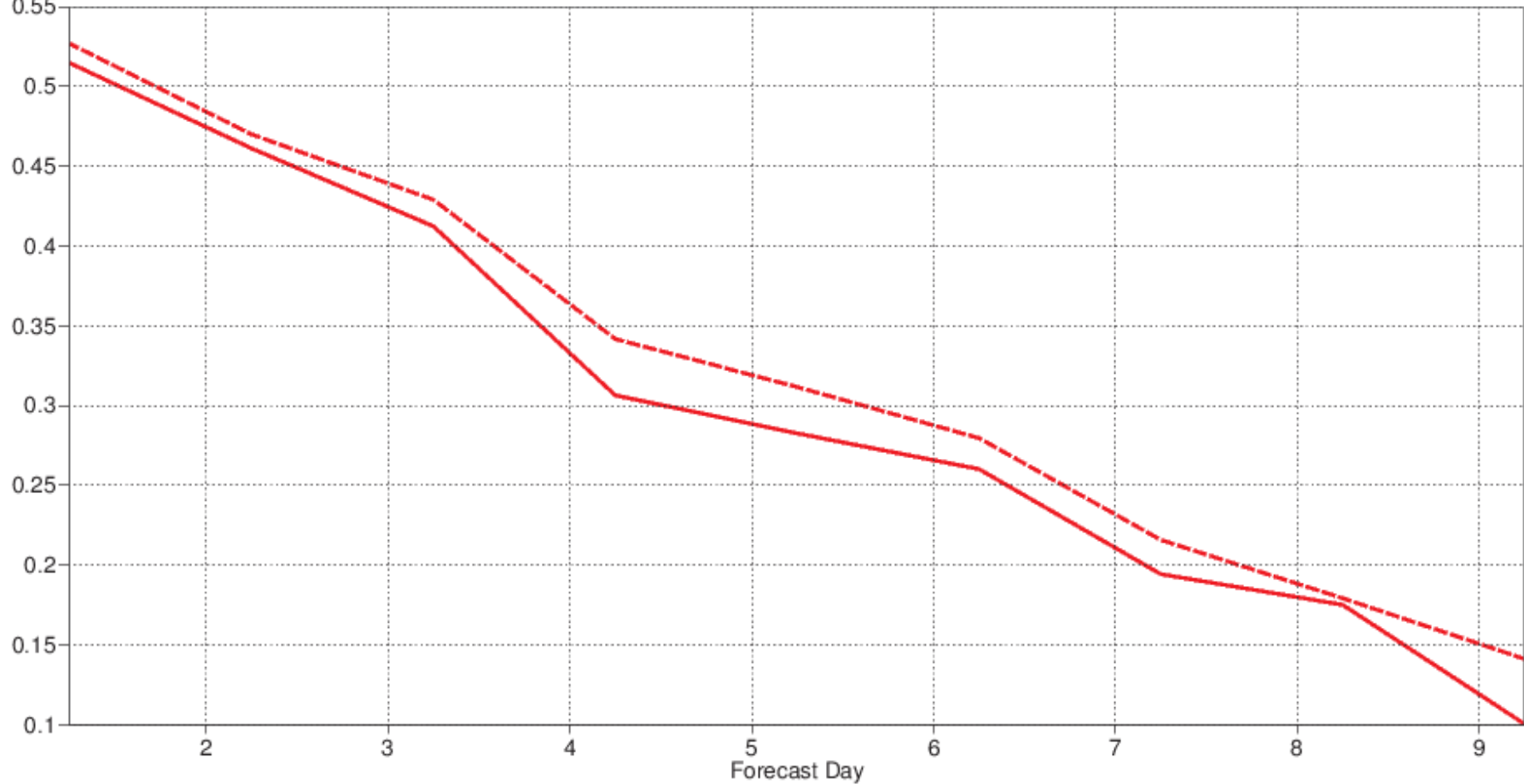
total precipitation

Equitable threat score value >5.0

France

Date: 201512 to 201602

HDOBS\_France od oper | Mean method: fair



# HDOBS 41r1-v-41r2 for Turkey, DJF 2015/16

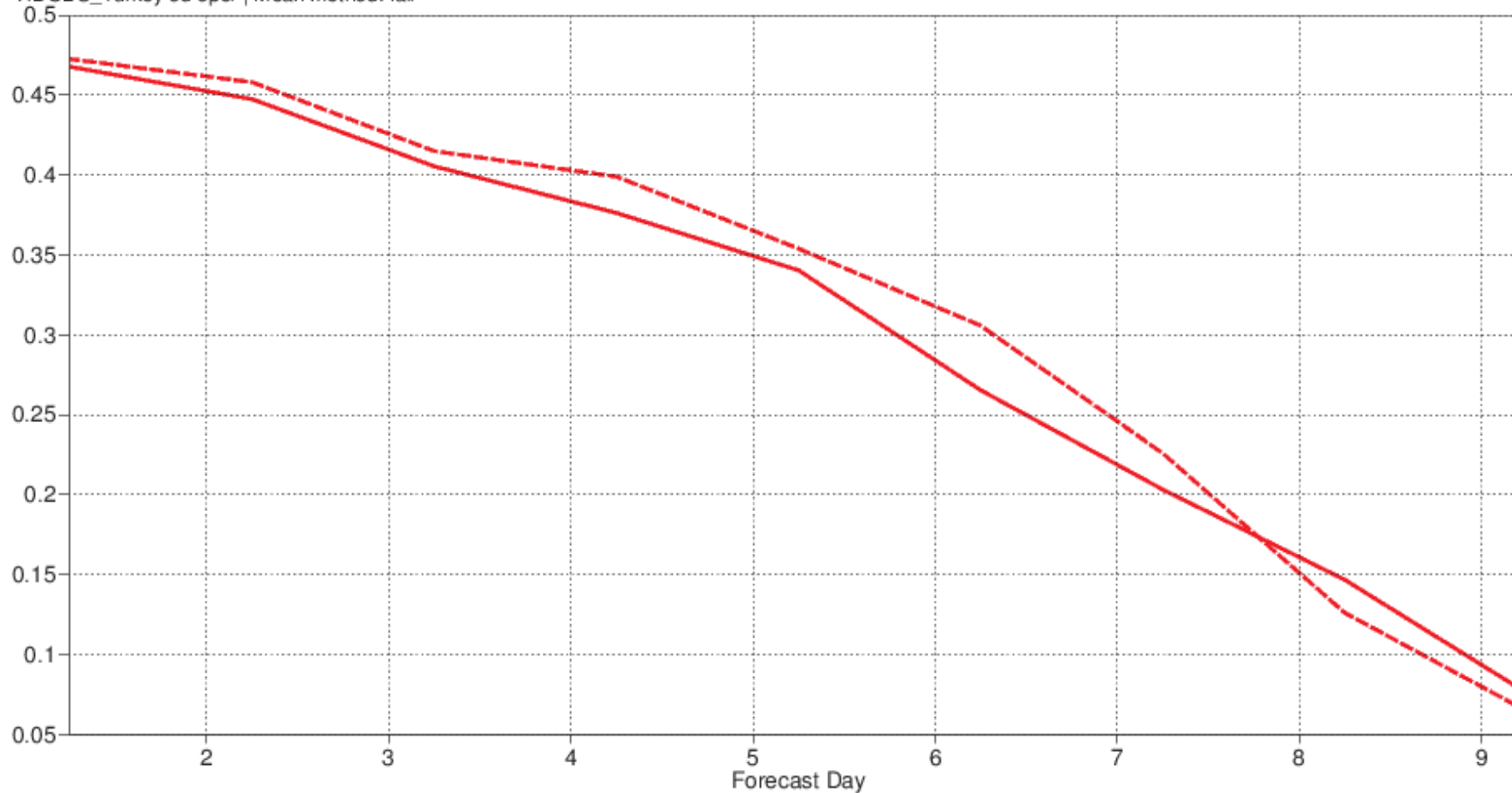
total precipitation

Equitable threat score value >5.0

Turkey

Date: 201512 to 201602

HDOBS\_Turkey od oper | Mean method: fair



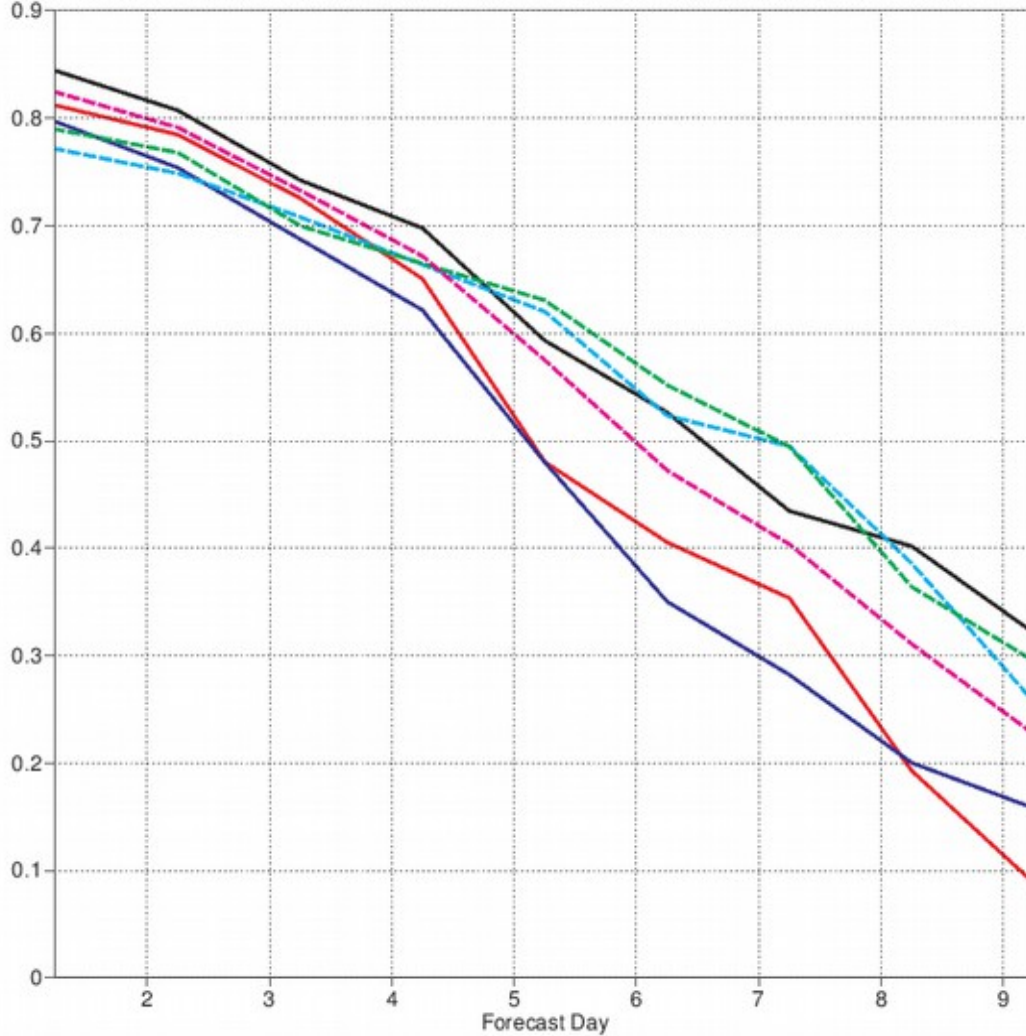


# HDOBS – Seasonality for Norway

total precipitation  
sedi value >5.0  
Norway

HDOBS\_Norway od oper 0001 | Mean method: standard

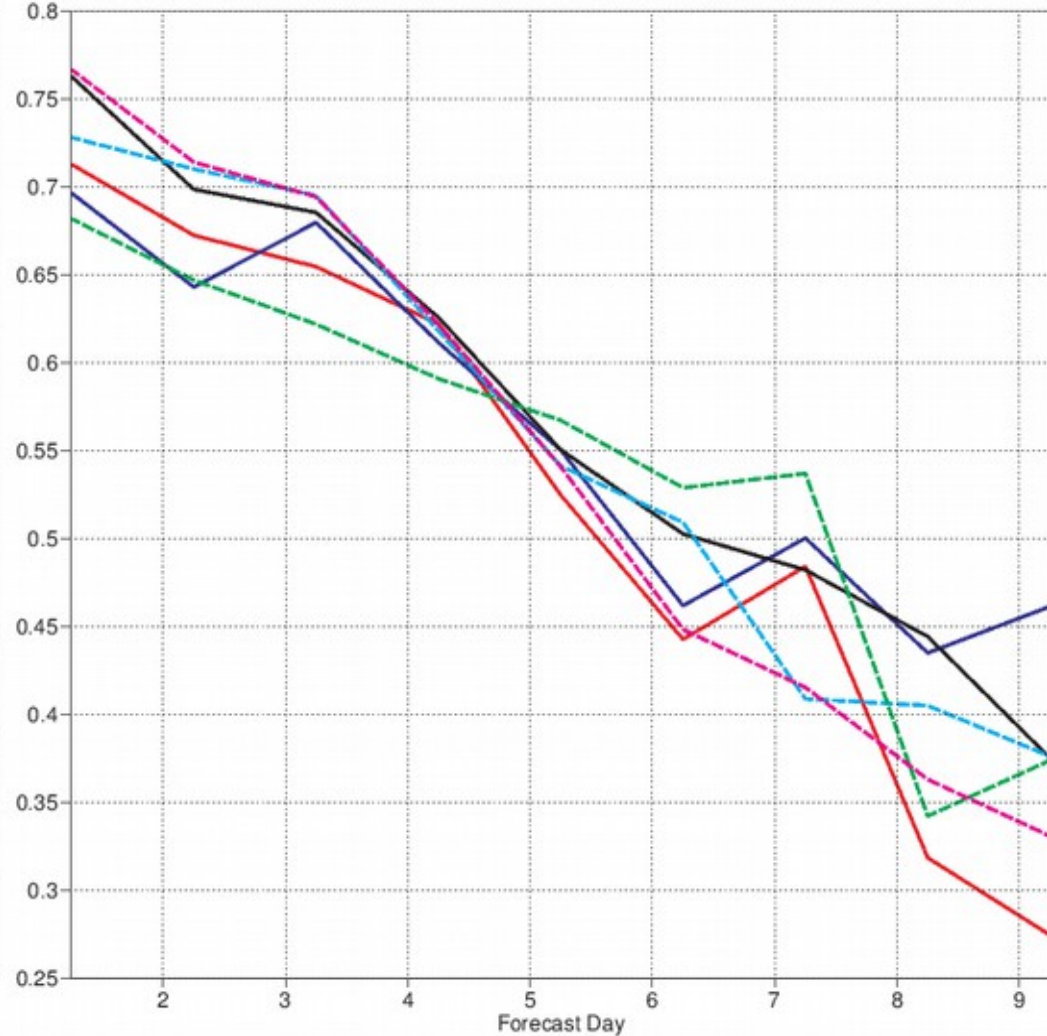
— SON15      - - - Nov14-Feb16  
— JJA15      - - - DJF15/16  
— MAM15      - - - DJF14/15



total precipitation  
sedi value >20.0  
Norway

HDOBS\_Norway od oper 0001 | Mean method: standard

— SON15      - - - Nov14-Feb16  
— JJA15      - - - DJF15/16  
— MAM15      - - - DJF14/15







# WGNE QPF Verifications over Japan Dec 2014–Nov 2015

JMA  
WGNE-31

# Data and Verification Method

## Verification grid

80 km×80 km

## Converting method

Simple average or interpolation

## Reference data (Observations)

Amount of precipitation observed by rain gauges

## Verified data (QPFs data)

See next slide

## Error bars

Estimated by bootstrap method  
with 95% confidence intervals

## Verification method

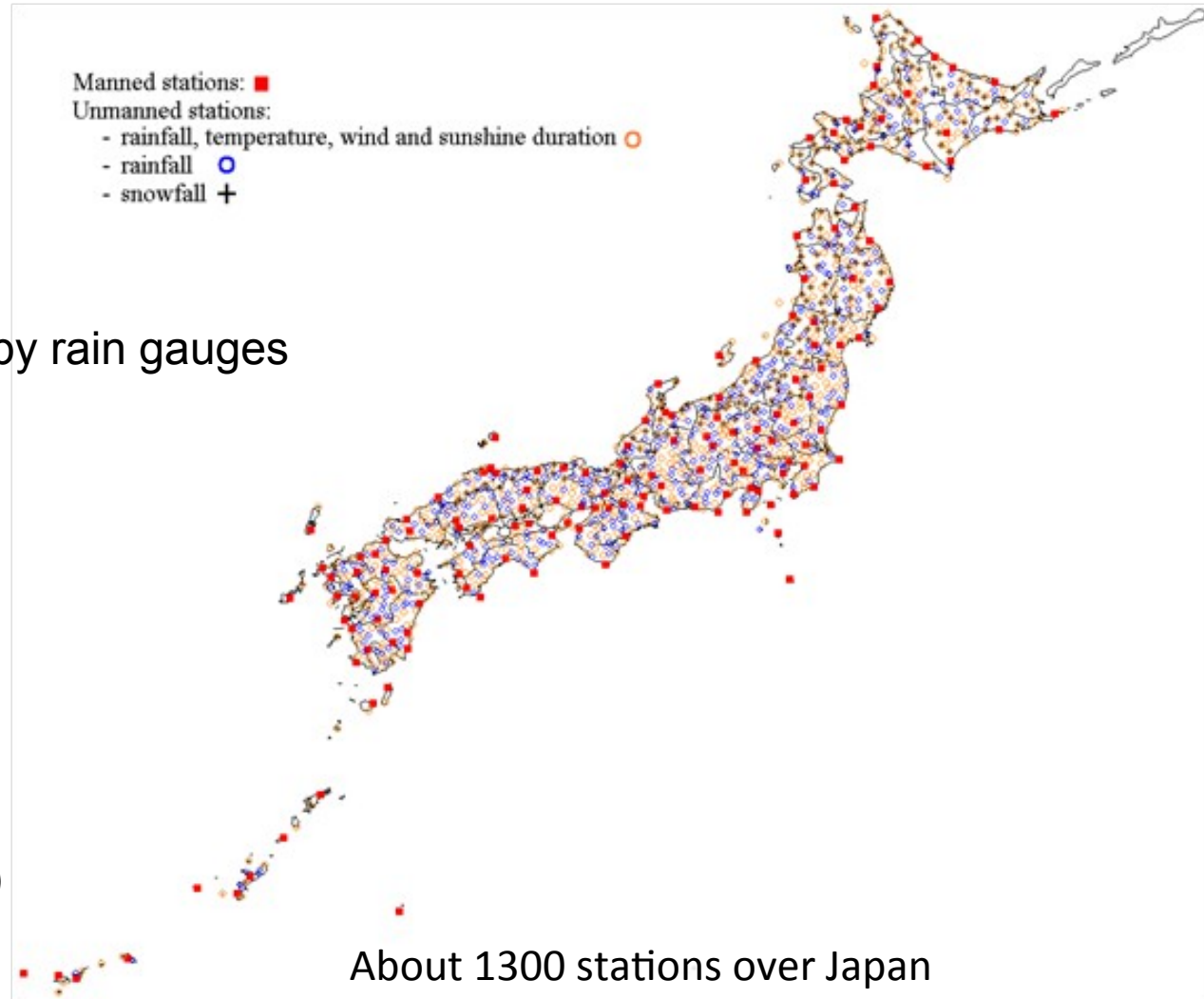
Equitable Thread Score (ETS)

Extremal Dependency Index (EDI)

Bias Score (BI, Optional)

Hit Rate (HR, Optional)

False Alarm Rate (FAR, Optional)



About 1300 stations over Japan

Average distance (among stations): ~17 km

# Verification with 80 km×80 km grid

NWP Center	horizontal resolution of verified data (degree)	forecast time (hour)	converting method in 80 km verification
BoM	0.5625 × 0.375	6, 12, ..., 144	average
CMC	1.00×1.00	6, 12, ..., 120	interpolation
DWD	0.25×0.25	6, 12, ..., 174	average
ECMWF	0.50×0.50	6, 12, ..., 72	average
NCEP	1.00×1.00 (*1) 0.50×0.50	6, 12, ..., 84	interpolation average
UKMO	0.234×0.156	6, 12, ..., 96	average
JMA	0.25×0.25 (GSM[*2]) 5 km×5 km (MSM[*3])	6, 12, ..., 84 3, 6, ..., 39	average average
Observation	Corresponding to 17 km×17 km	—	average

(\*1) before 2015/01/14

(\*2) global model

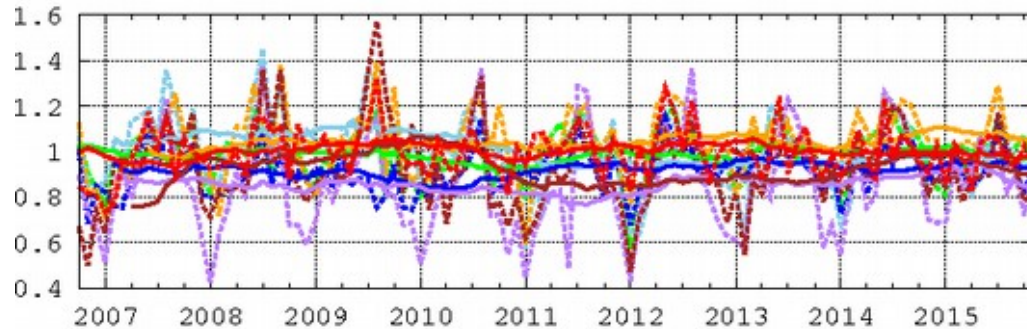
(\*3) regional model



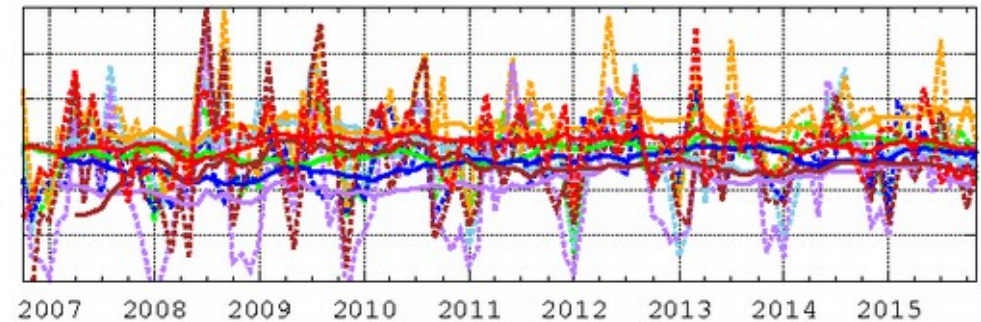
# Time series from late 2006

JMA — ECMWF — NCEP — UKMO — CMC — DWD — BoM

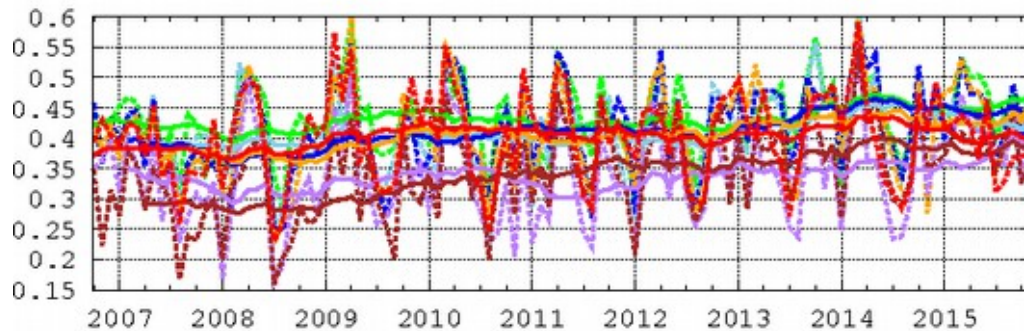
Bias Score 10mm/24hr FT24-48



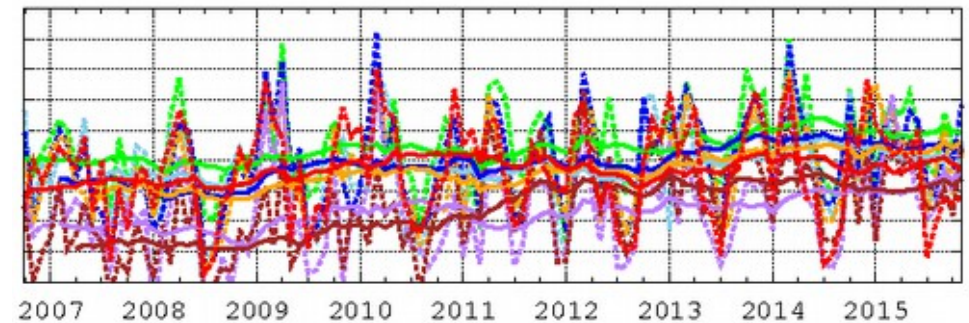
Bias Score 10mm/24hr FT48-72



Equitable Threat Score 10mm/24hr FT24-48

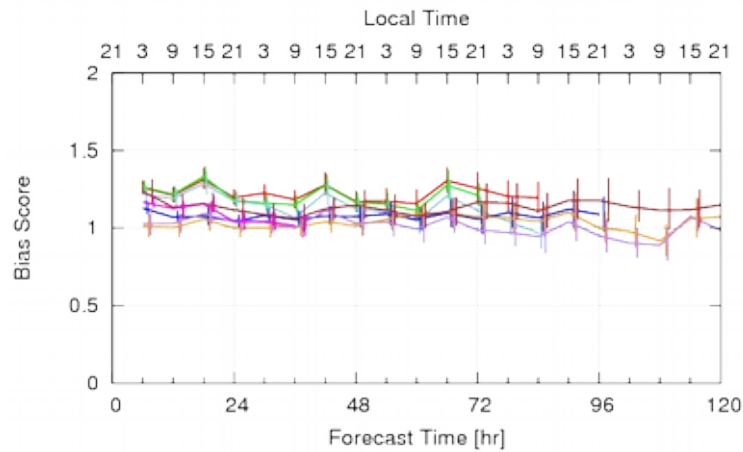


Equitable Threat Score 10mm/24hr FT48-72

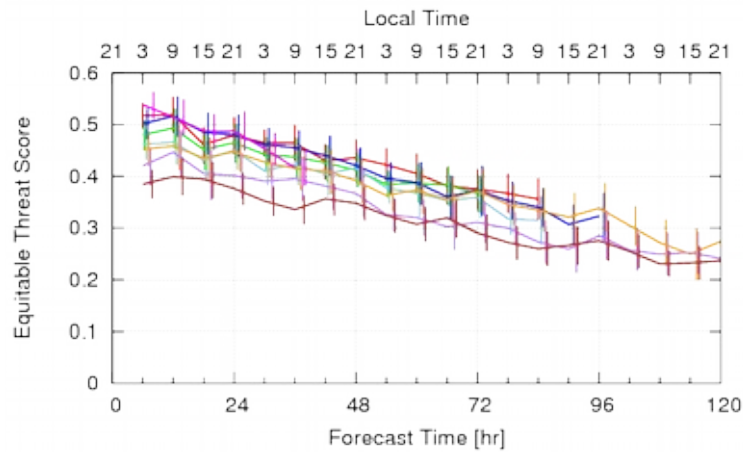


○ETS of DWD increases in 2015.

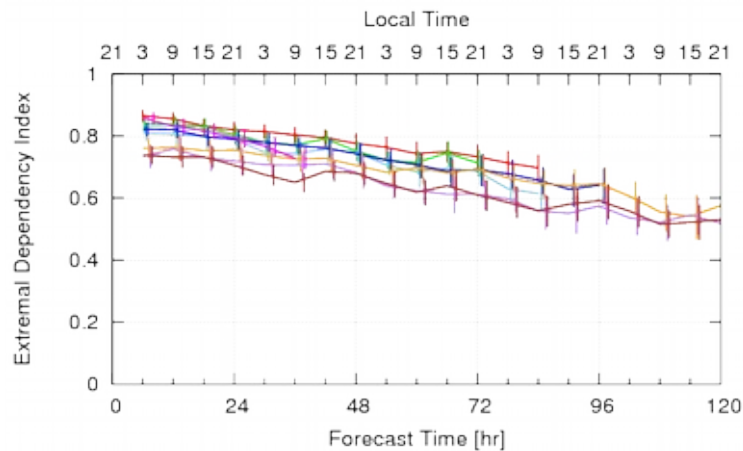
NOTE: Solid lines represent moving-average (12 months).



Equitable Threat Score: 1.0mm/6h 2014/12-2015/02



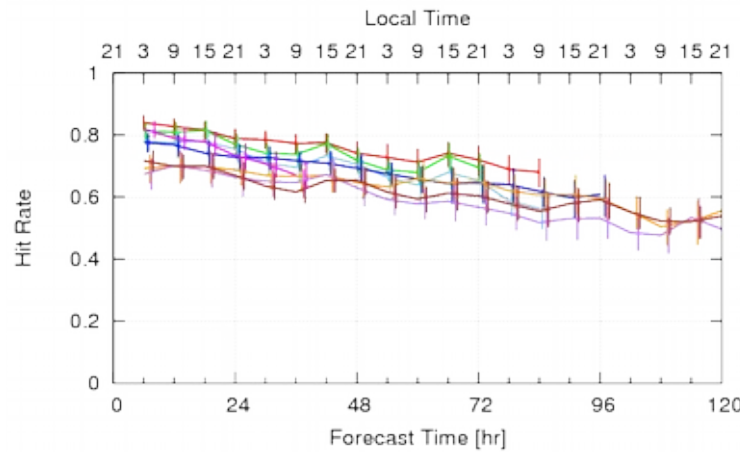
Extremal Dependency Index: 1.0mm/6h 2014/12-2015/02



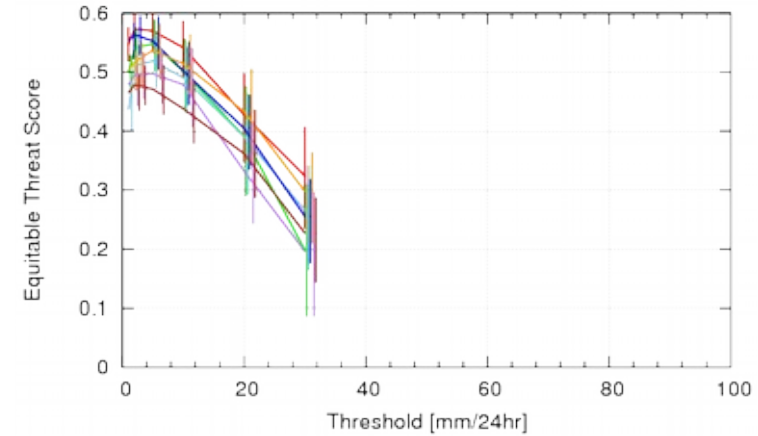
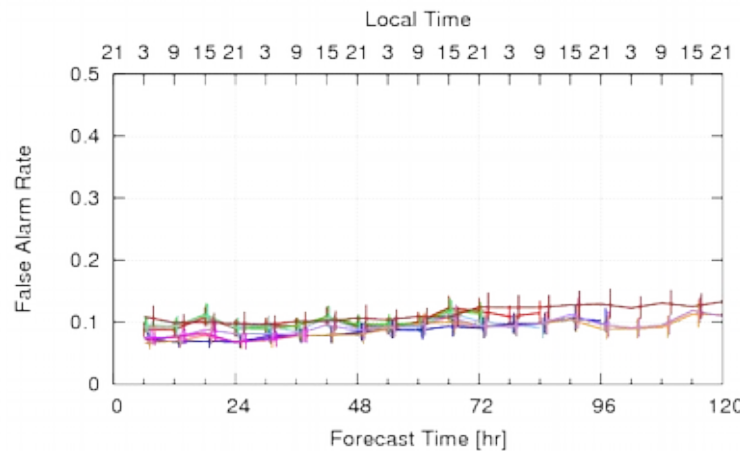
# 2014DJF

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

Hit Rate: 1.0mm/6h 2014/12-2015/02



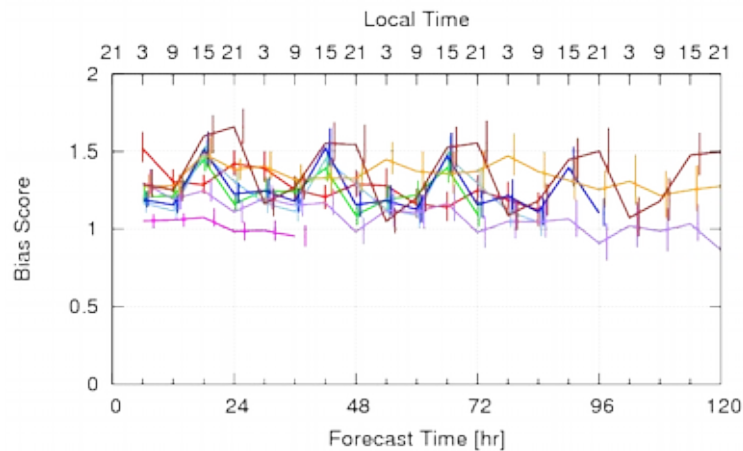
False Alarm Rate: 1.0mm/6h 2014/12-2015/02



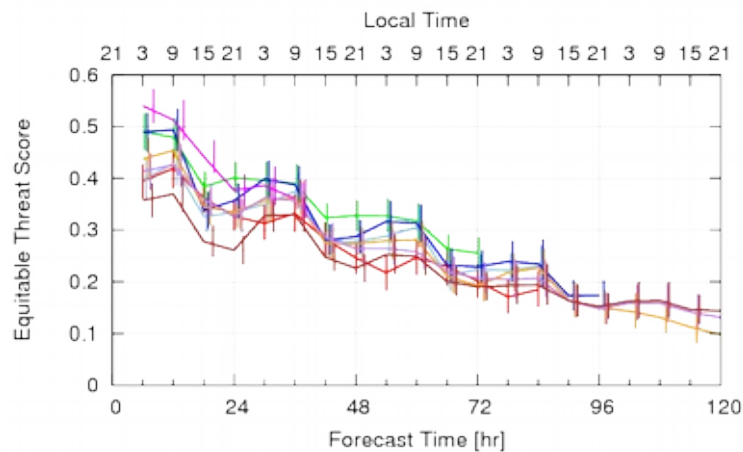
- At 1mm/6hr threshold, JMA performs better in ETS, EDI, and HR, but has large BI and FAR.
- ECMWF performs better in ETS at low thresholds.
- CMC performs better in ETS at high threshold.



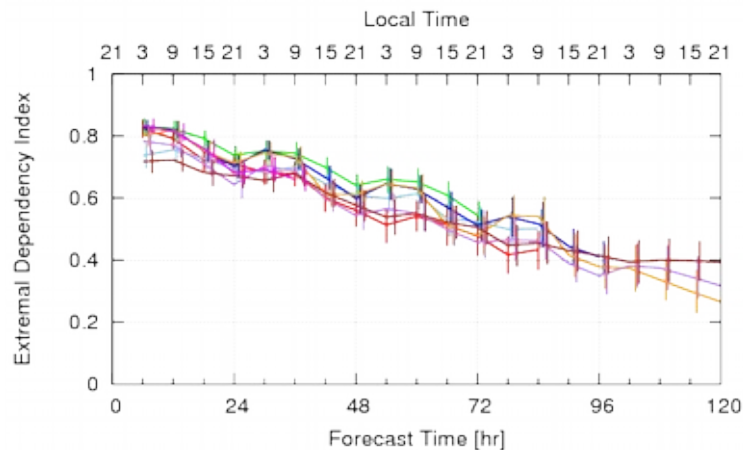
Bias Score: 1.0mm/6h 2015/06-2015/08



Equitable Threat Score: 1.0mm/6h 2015/06-2015/08



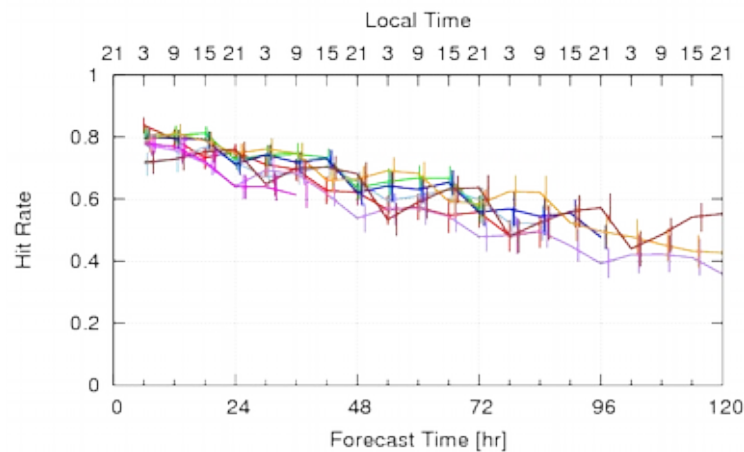
Extremal Dependency Index: 1.0mm/6h 2015/06-2015/08



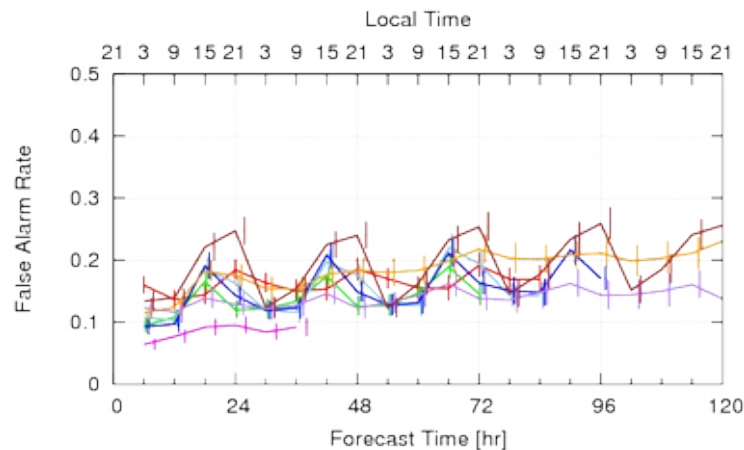
# 2015JJA

— JMA  
— ECMWF  
— NCEP  
— UKMO  
— CMC  
— DWD  
— BoM  
— MSM

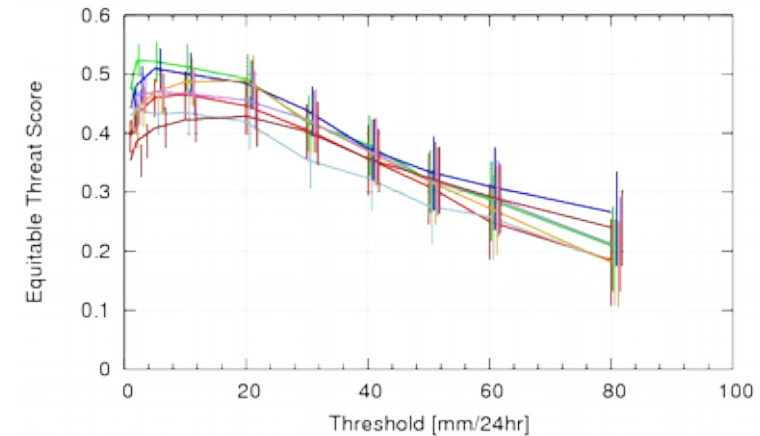
Hit Rate: 1.0mm/6h 2015/06-2015/08



False Alarm Rate: 1.0mm/6h 2015/06-2015/08



Equitable Threat Score: FT0-24 2015/06-2015/08



- In BI, diurnal cycle is seen in many centers (large BI at 15 local time).
- At 1 mm/6hr threshold, MSM performs quite better in BI.
- ECMWF performs better in ETS at thresholds below 50 mm/24hr.
- UKMO performs better in ETS at any thresholds.

# Backup Slides

# Data and Verification Method

## Verification grid

80 km×80 km

## Converting method

Simple average or interpolation

## Reference data (Observations)

Amount of precipitation observed by rain gauges

## Verified data (QPFs data)

See next slide

## Error bars

Estimated by bootstrap method  
with 95% confidence intervals

## Verification method

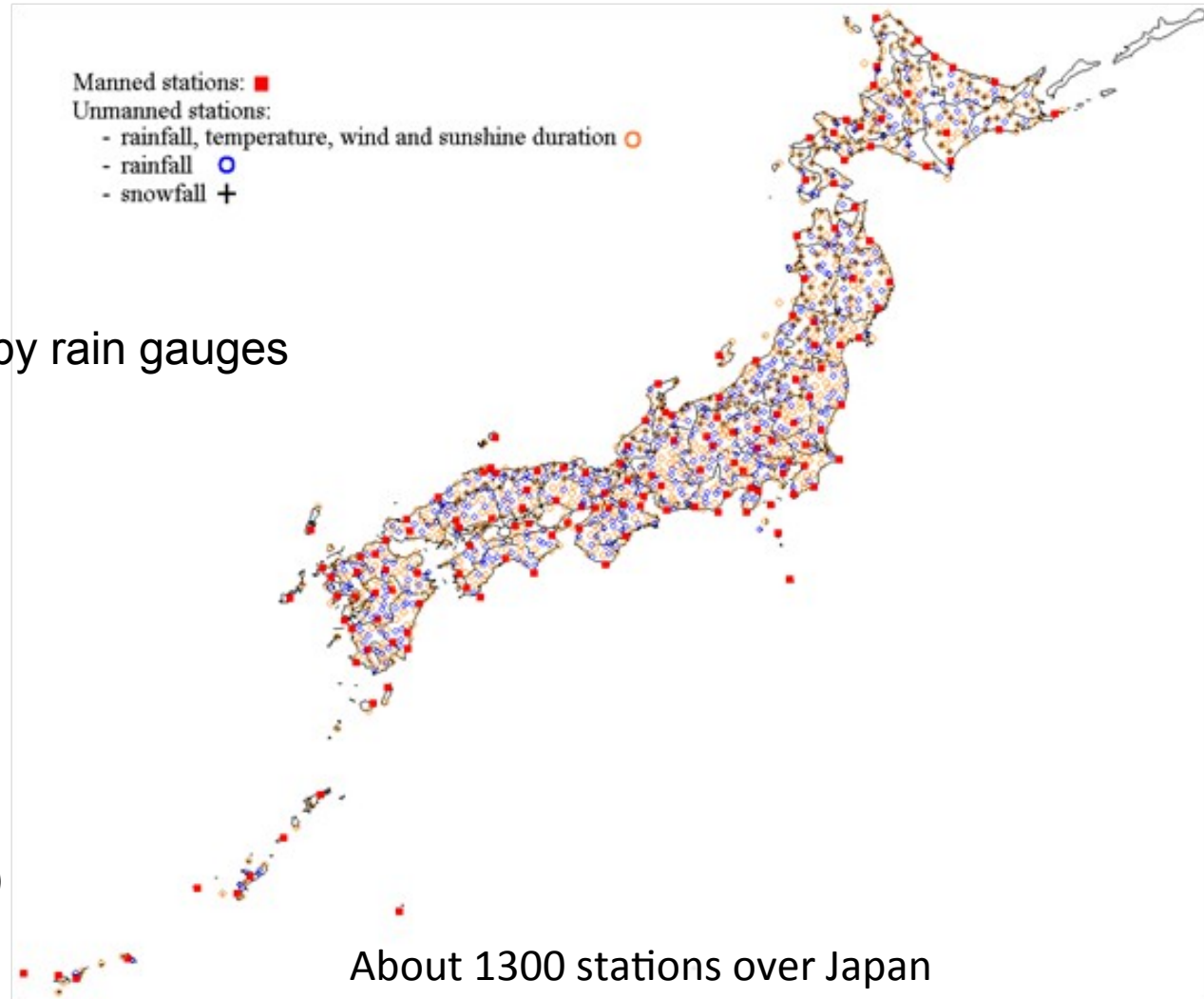
Equitable Thread Score (ETS)

Extremal Dependency Index (EDI)

Bias Score (BI, Optional)

Hit Rate (HR, Optional)

False Alarm Rate (FAR, Optional)



About 1300 stations over Japan

Average distance (among stations): ~17 km

# Verification with 80 km×80 km grid

NWP Center	horizontal resolution of verified data (degree)	forecast time (hour)	converting method in 80 km verification
BoM	0.5625 × 0.375	6, 12, ..., 144	average
CMC	1.00×1.00	6, 12, ..., 120	interpolation
DWD	0.25×0.25	6, 12, ..., 174	average
ECMWF	0.50×0.50	6, 12, ..., 72	average
NCEP	1.00×1.00 (*1) 0.50×0.50	6, 12, ..., 84	interpolation average
UKMO	0.234×0.156	6, 12, ..., 96	average
JMA	0.25×0.25 (GSM[*2]) 5 km×5 km (MSM[*3])	6, 12, ..., 84 3, 6, ..., 39	average average
Observation	Corresponding to 17 km×17 km	—	average

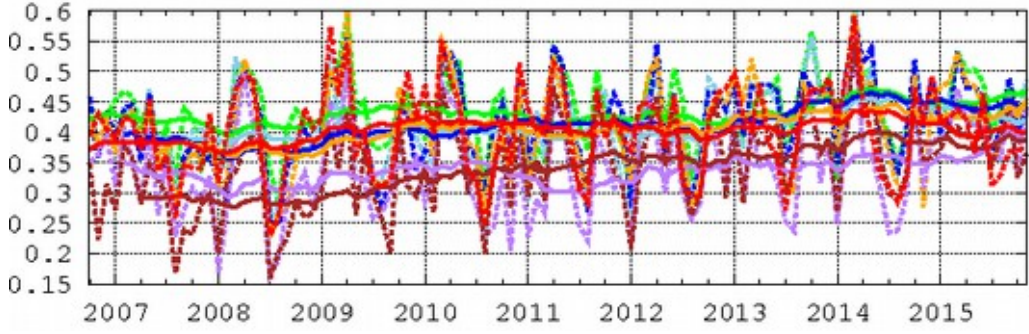
(\*1) before 2015/01/14

(\*2) global model

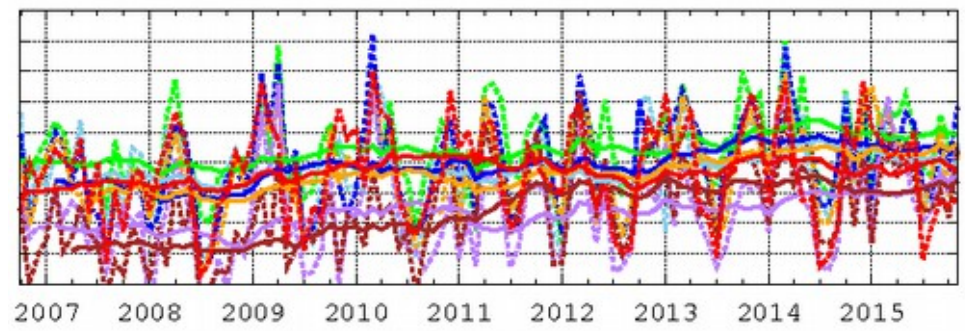
(\*3) regional model

JMA — ECMWF — NCEP — UKMO — CMC — DWD — BoM

Equitable Threat Score 10mm/24hr FT24-48

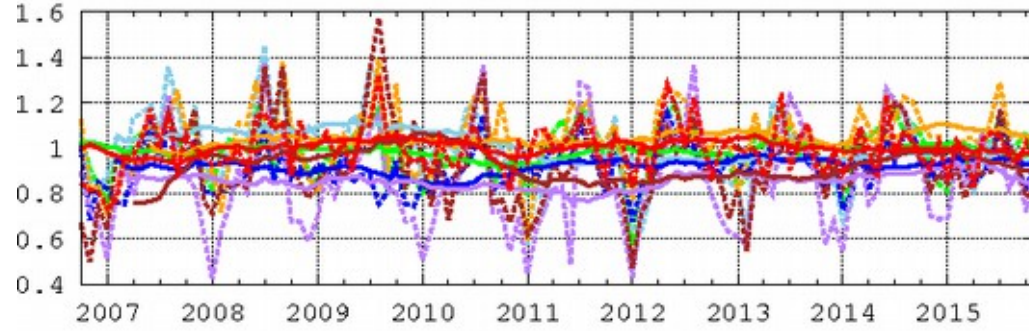


Equitable Threat Score 10mm/24hr FT48-72

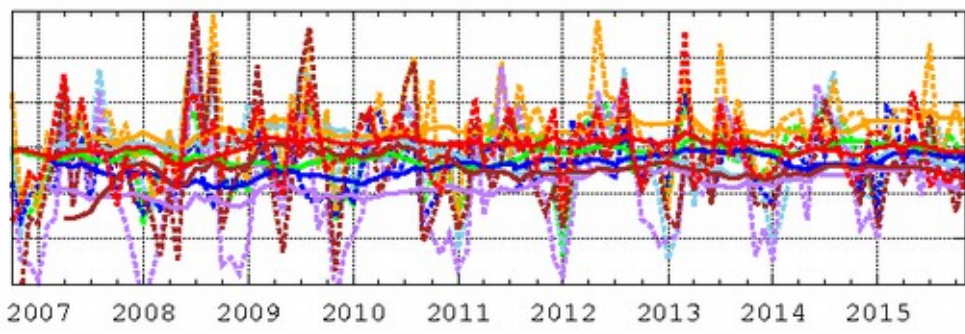


- ETS of DWD increases in 2015.
- ETS of JMA in 2015 is worse than that in 2014.

Bias Score 10mm/24hr FT24-48



Bias Score 10mm/24hr FT48-72



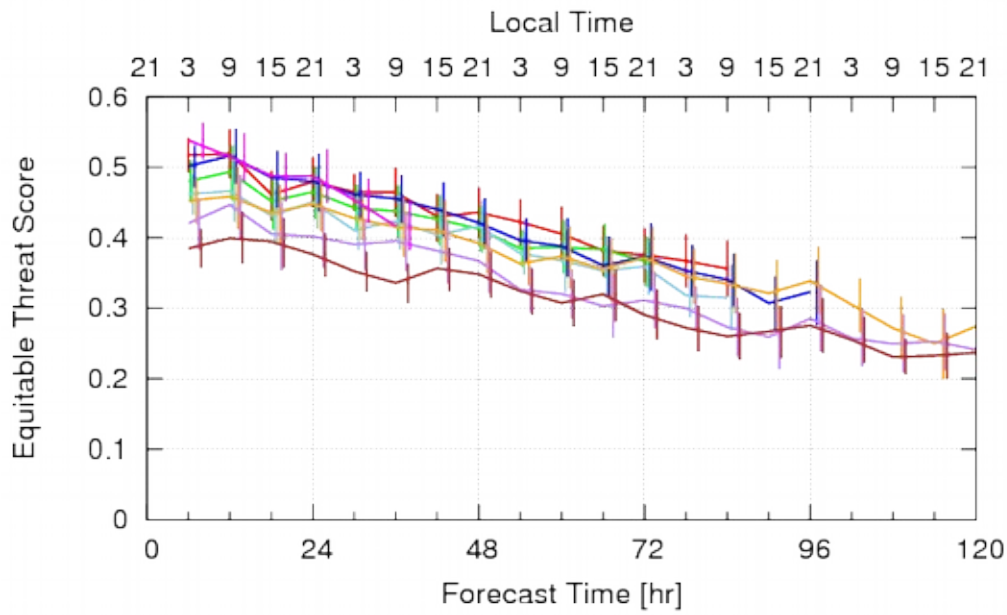
NOTE: Solid lines represent moving-average (12 months).



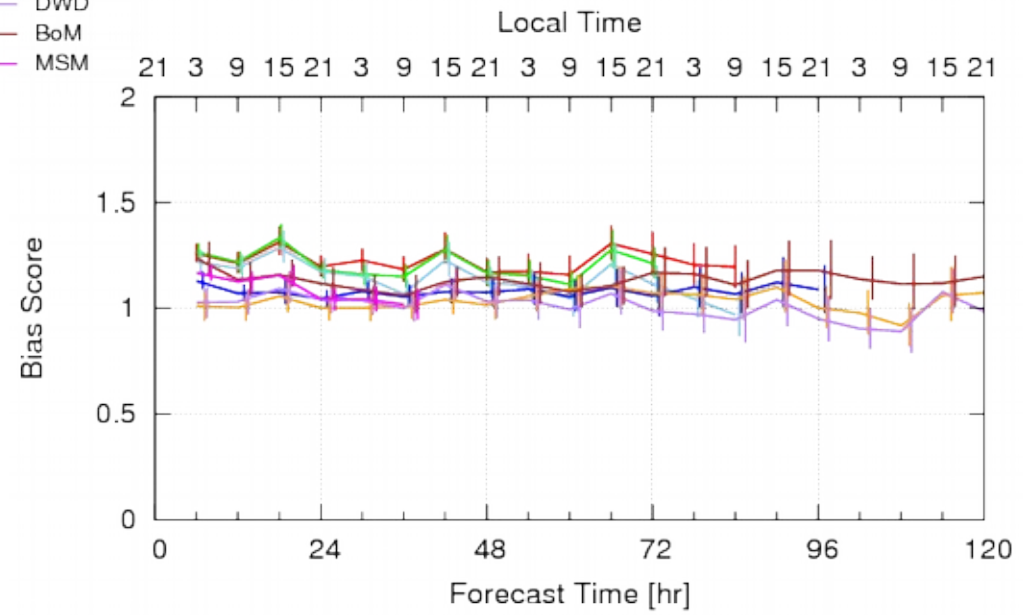
2014DJF

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

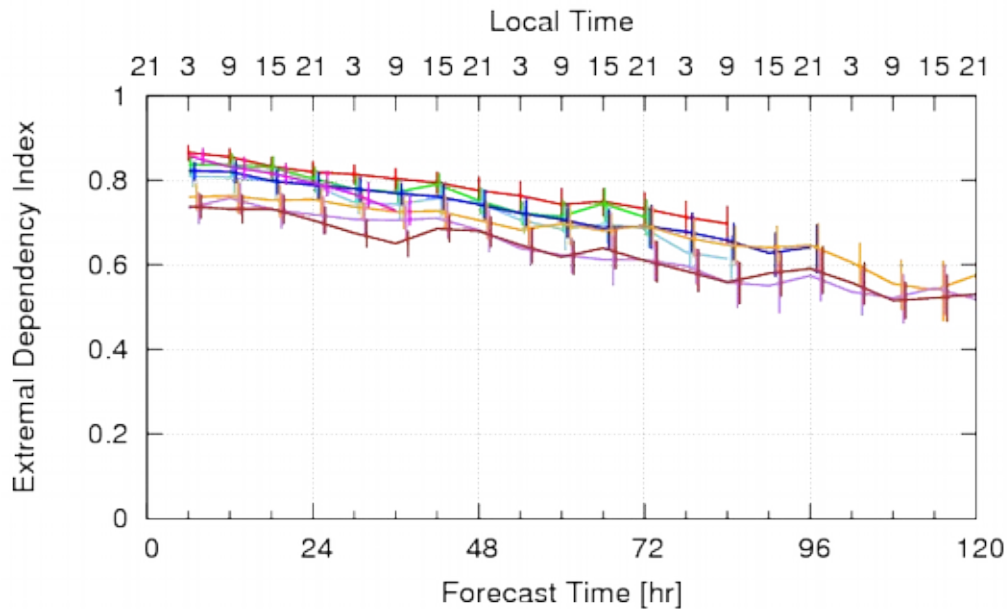
Equitable Threat Score: 1.0mm/6h 2014/12-2015/02



Bias Score: 1.0mm/6h 2014/12-2015/02

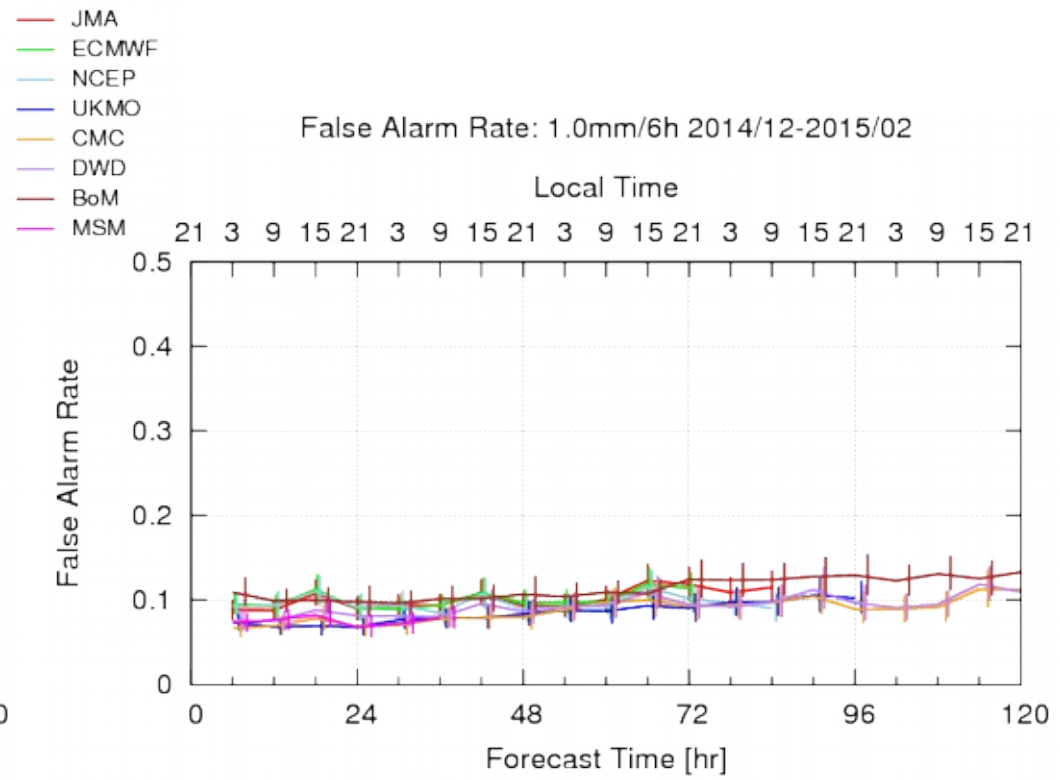
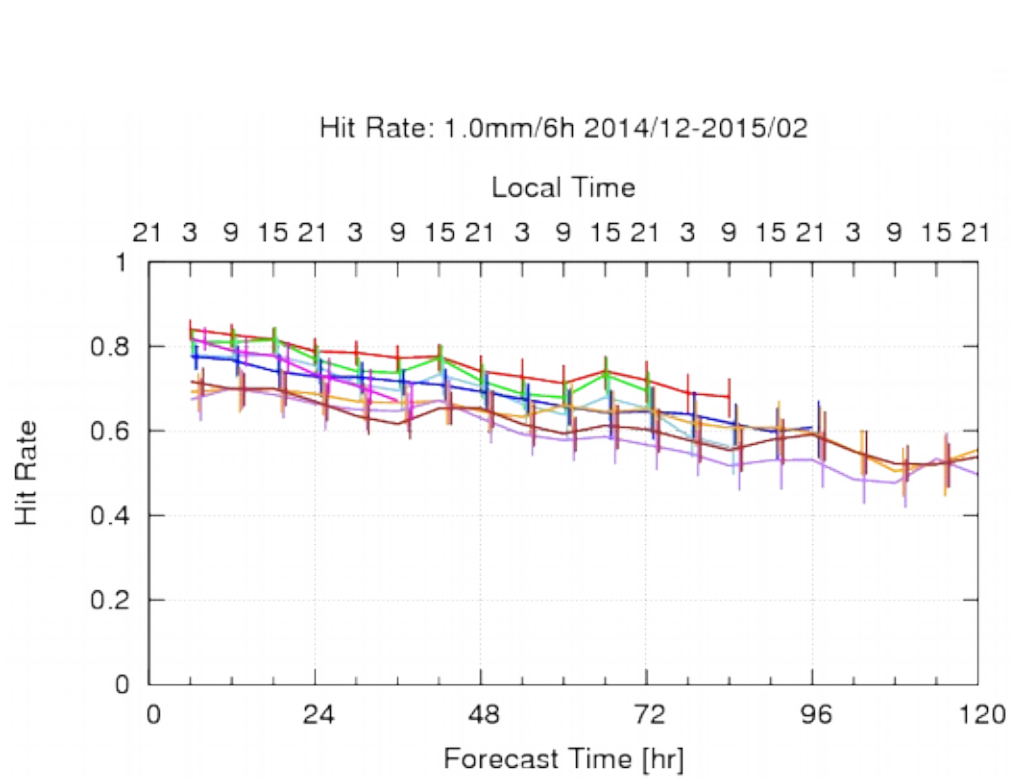


Extremal Dependency Index: 1.0mm/6h 2014/12-2015/02



○ JMA performs better in ETS and EDI, but has large BI.

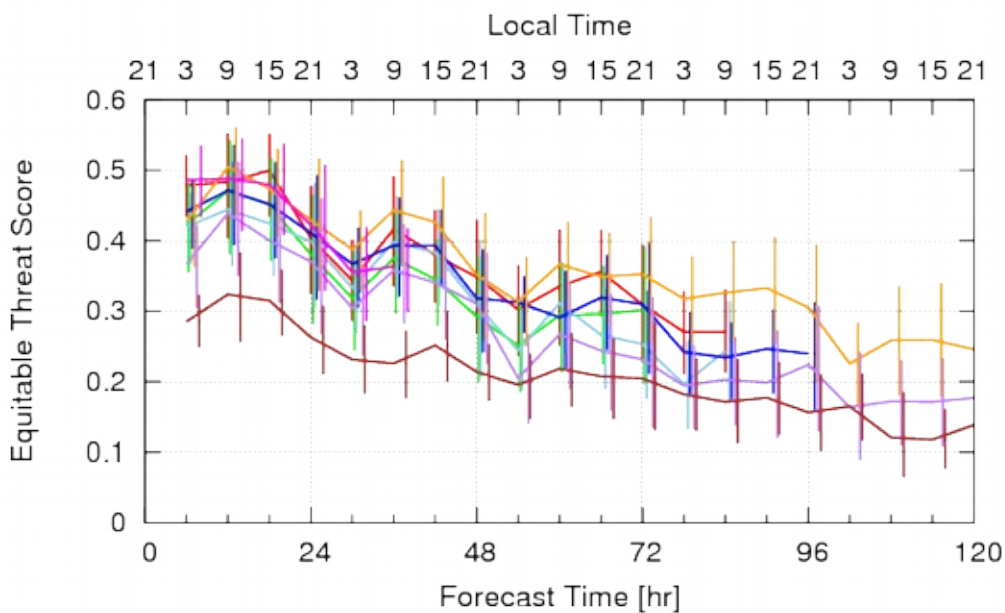
NOTE: Error bars are shifted slightly for clarification.



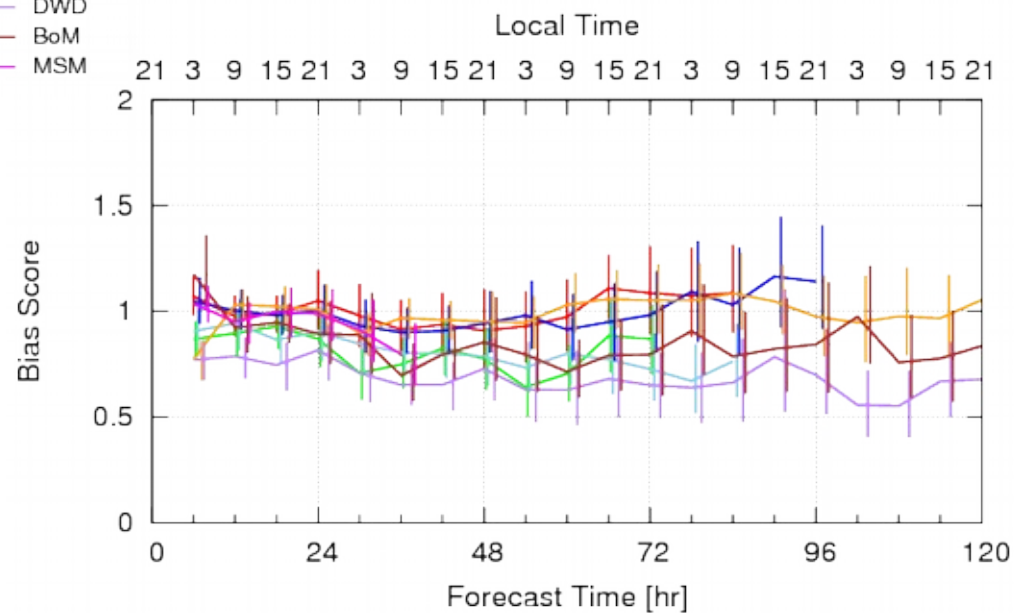
○ JMA performs better in HR, but has rather large FAR.

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

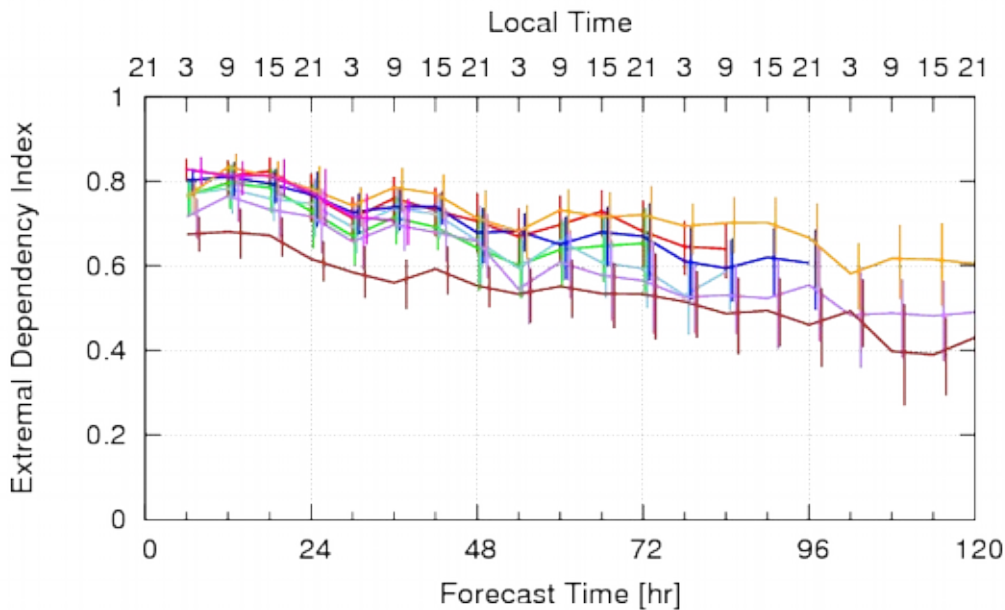
Equitable Threat Score: 5.0mm/6h 2014/12-2015/02



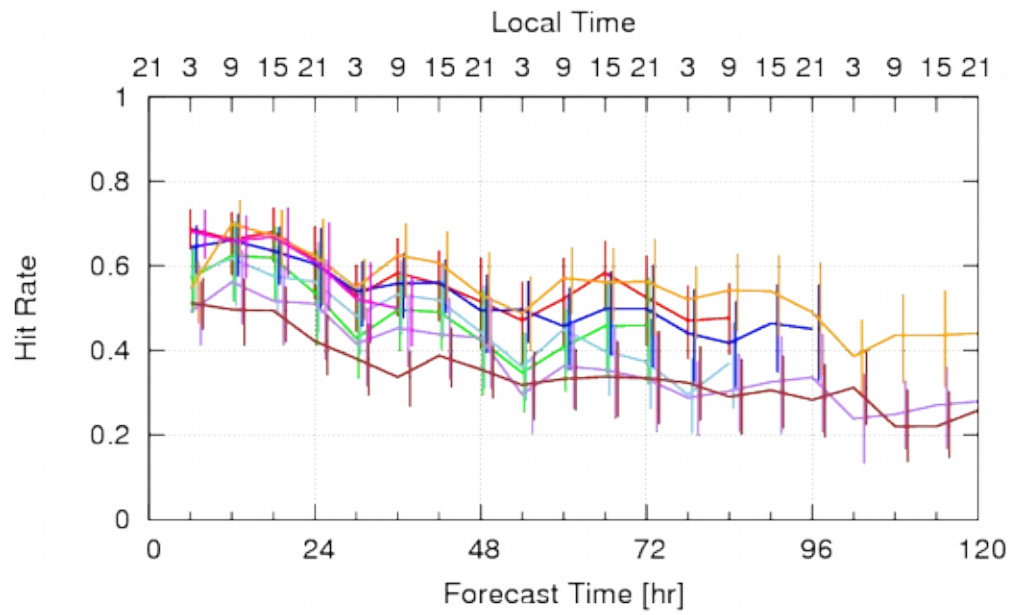
Bias Score: 5.0mm/6h 2014/12-2015/02



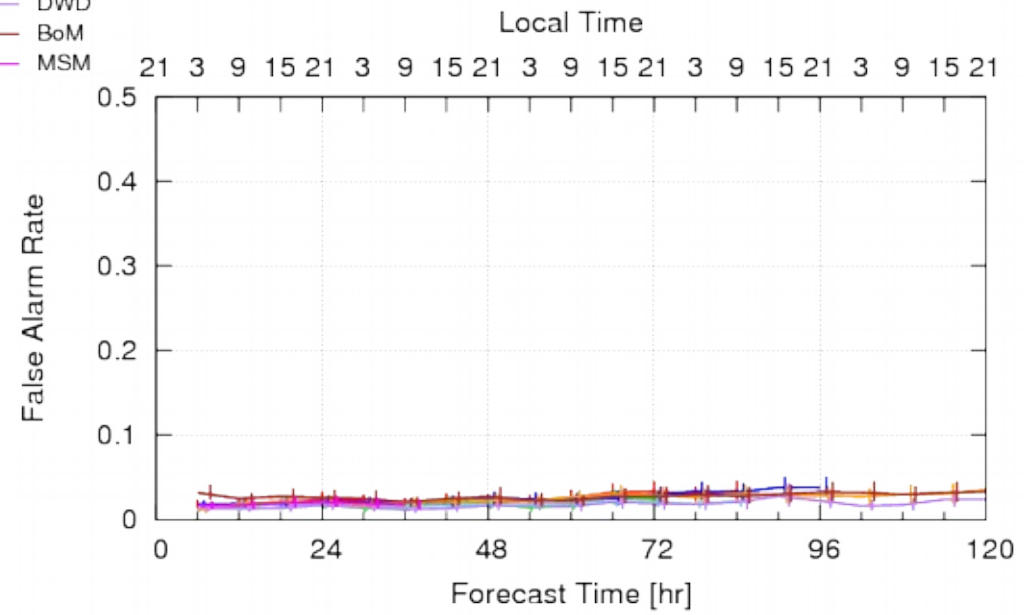
Extremal Dependency Index: 5.0mm/6h 2014/12-2015/02



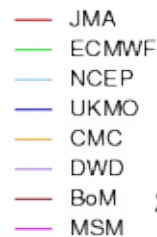
Hit Rate: 5.0mm/6h 2014/12-2015/02



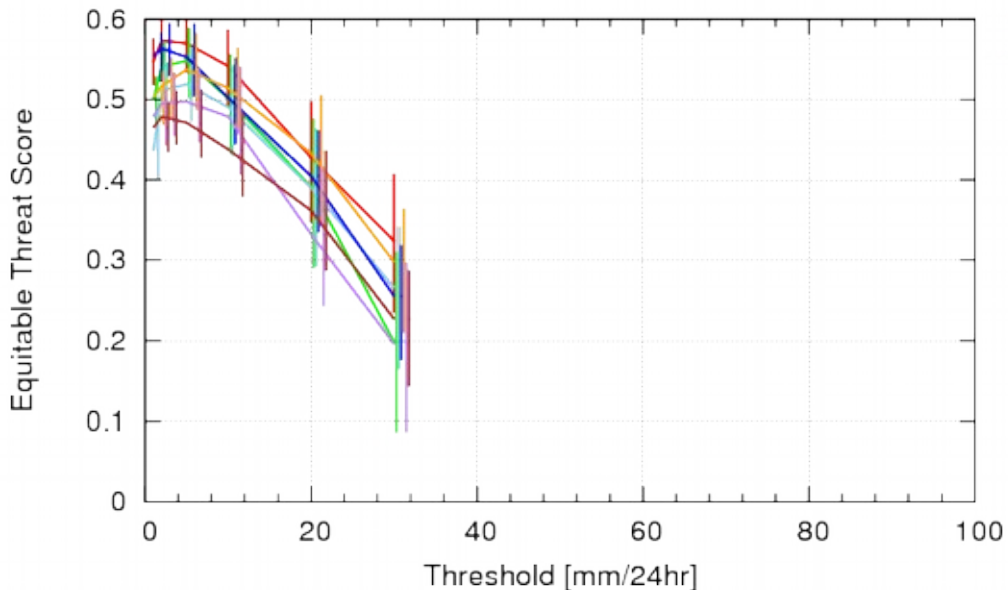
False Alarm Rate: 5.0mm/6h 2014/12-2015/02



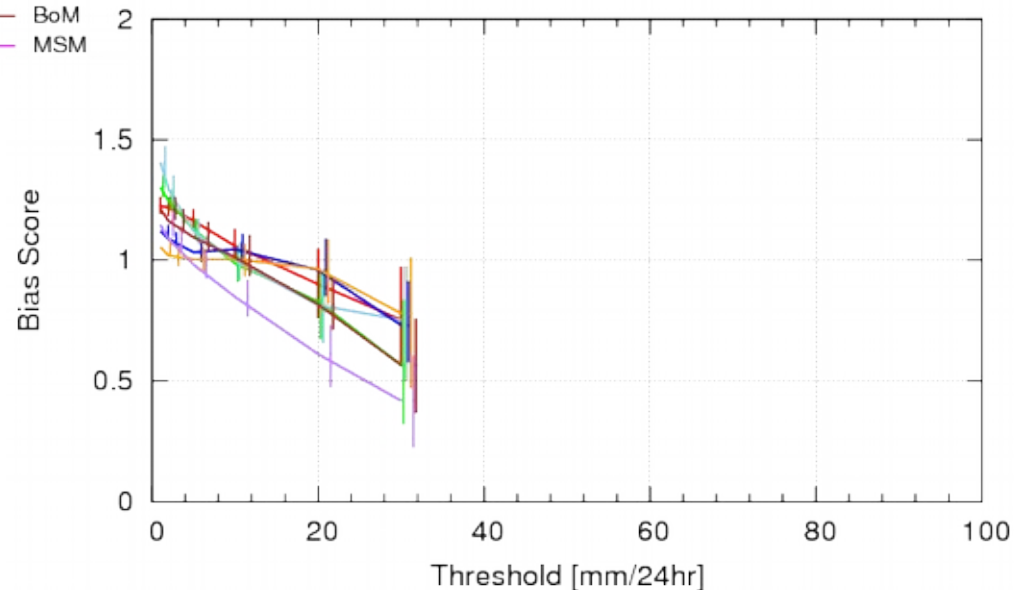




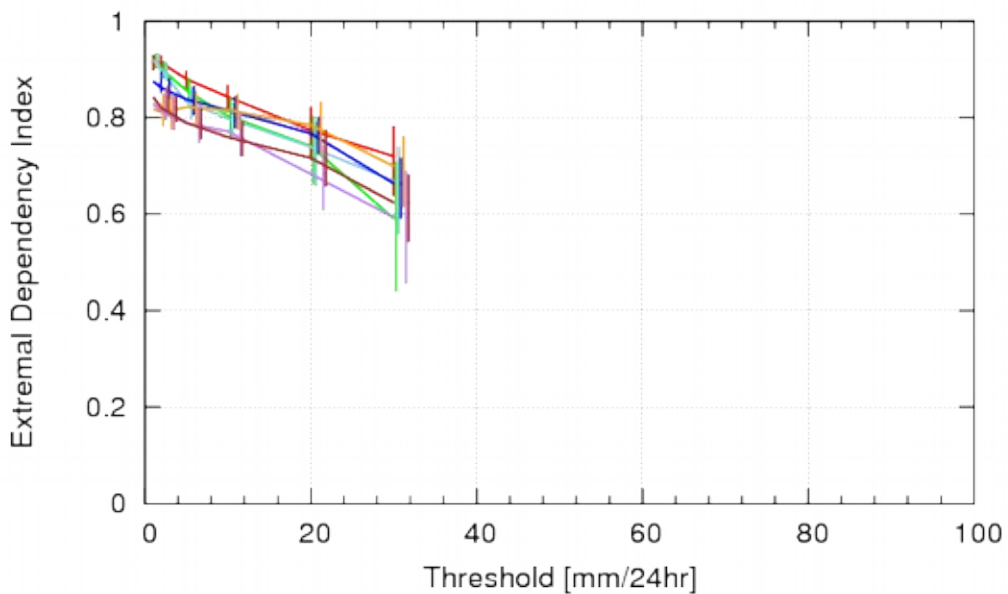
Equitable Threat Score: FT0-24 2014/12-2015/02



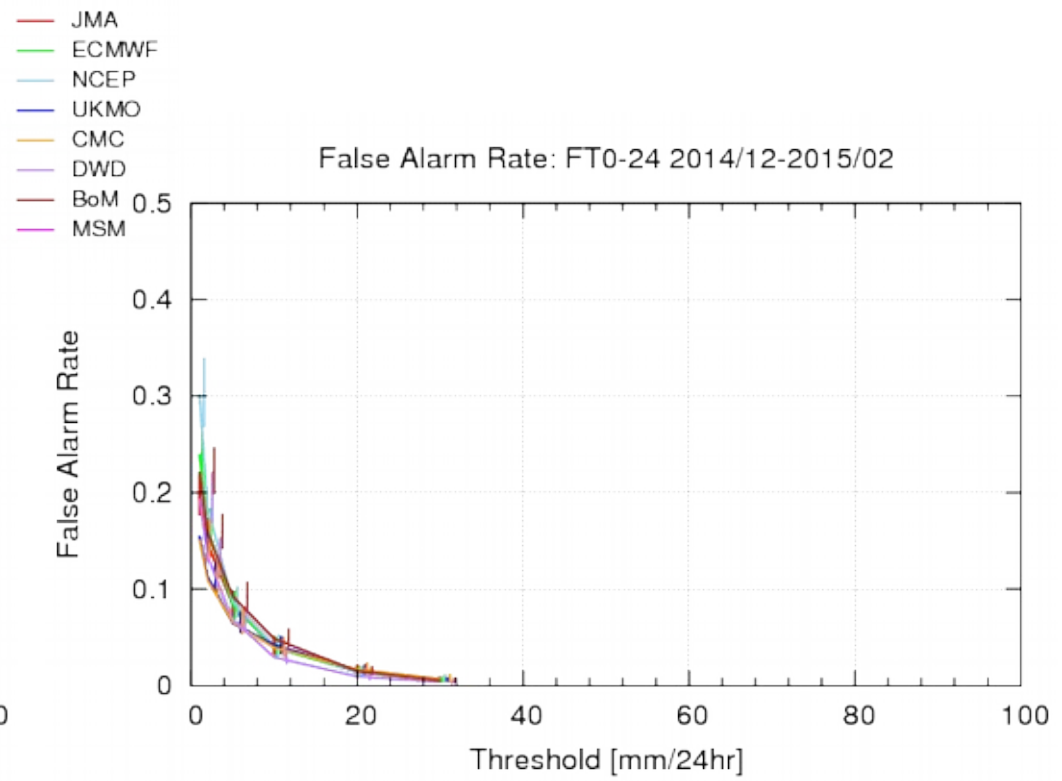
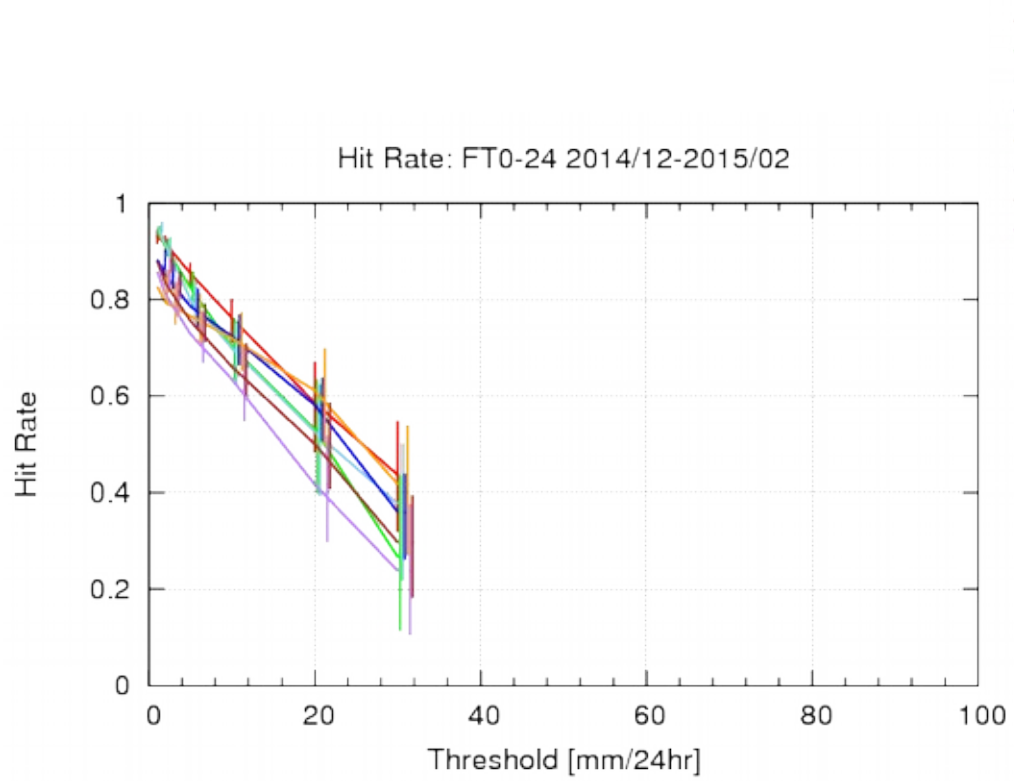
Bias Score: FT0-24 2014/12-2015/02



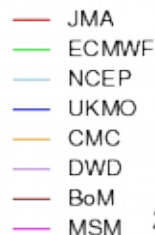
Extremal Dependency Index: FT0-24 2014/12-2015/02



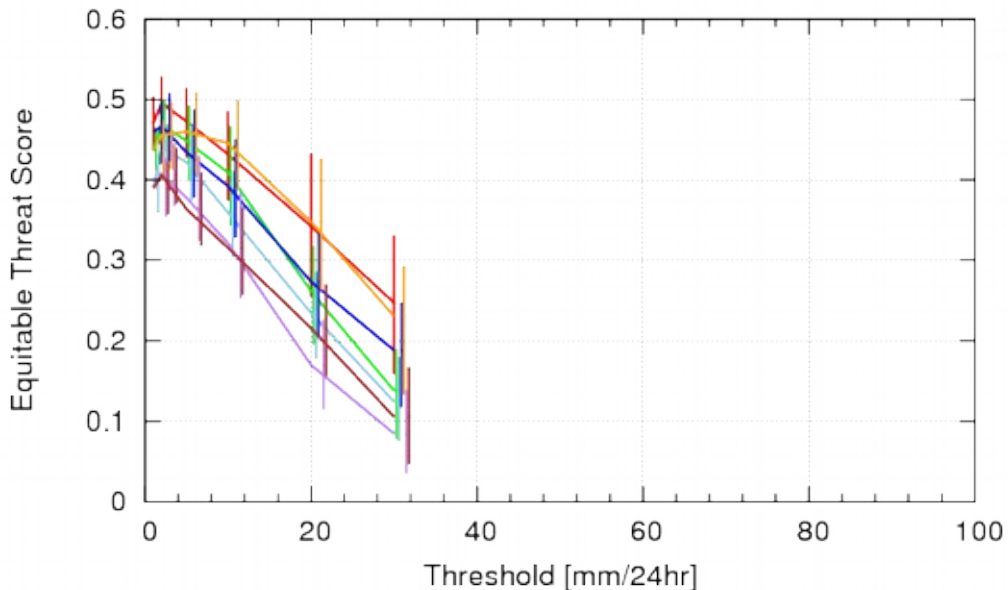
- CMC performs better in BI.
- JMA performs better in ETS and EDI.
- ECMWF performs better in ETS and EDI at the low threshold.
- CMC performs better in ETS and EDI at the high threshold.



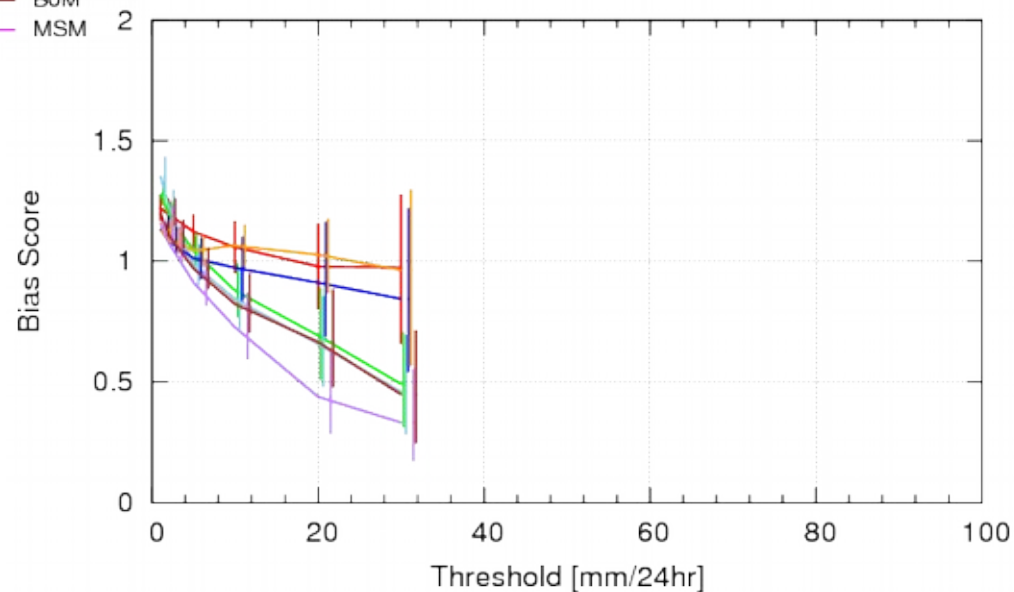
○ JMA performs better in HR.



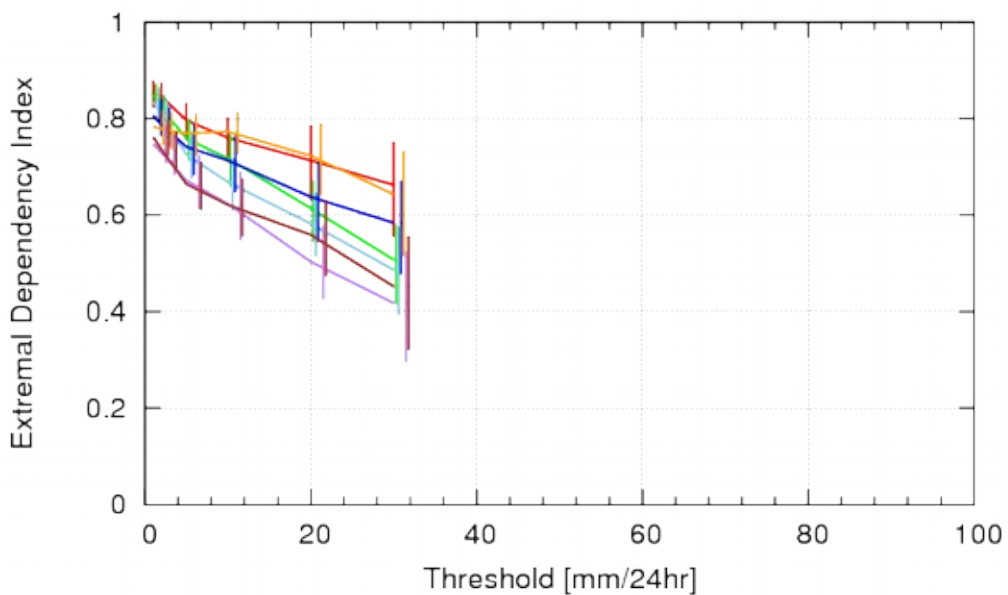
Equitable Threat Score: FT48-72 2014/12-2015/02



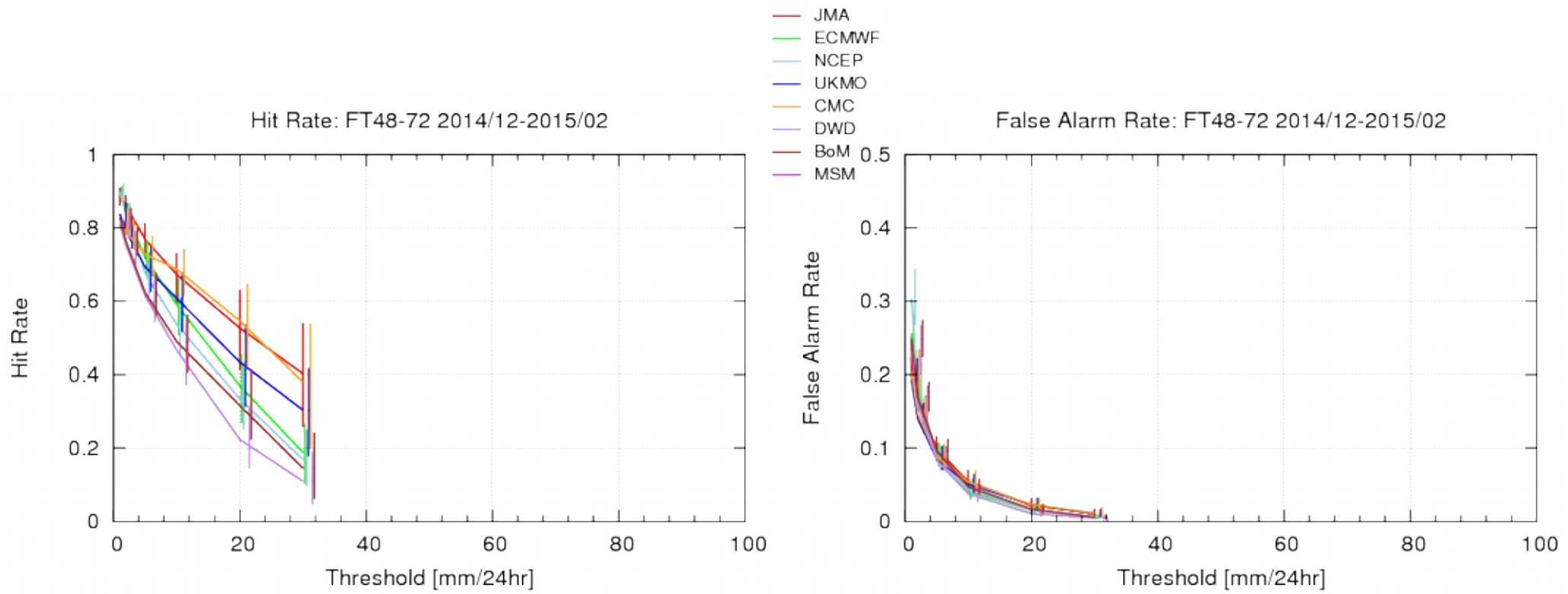
Bias Score: FT48-72 2014/12-2015/02



Extremal Dependency Index: FT48-72 2014/12-2015/02



○JMA performs better in ETS and EDI.  
 ○CMC performs better in ETS and EDI at the high threshold.



○ JMA performs better in HR.

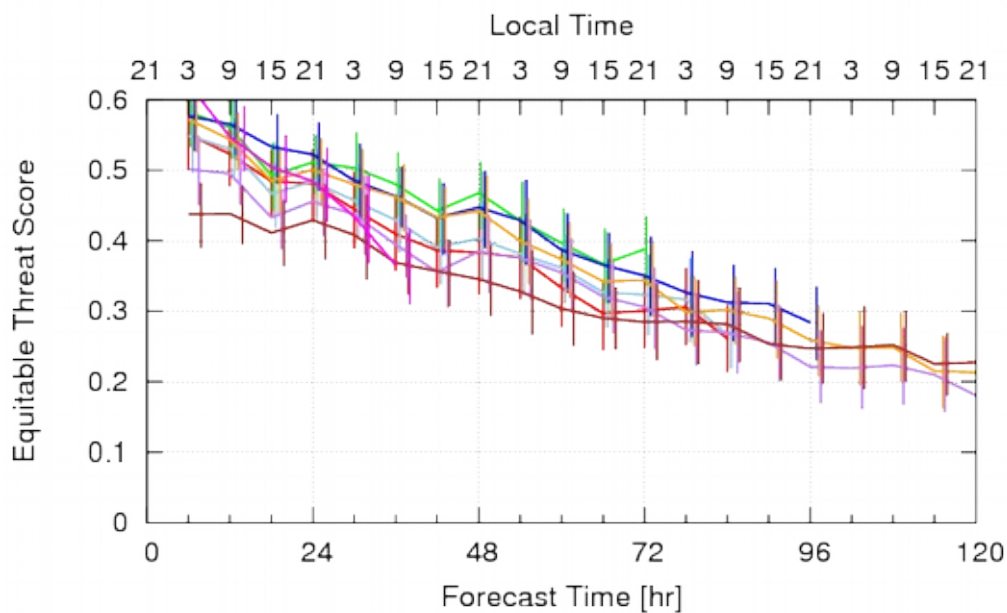
**2015MAM**



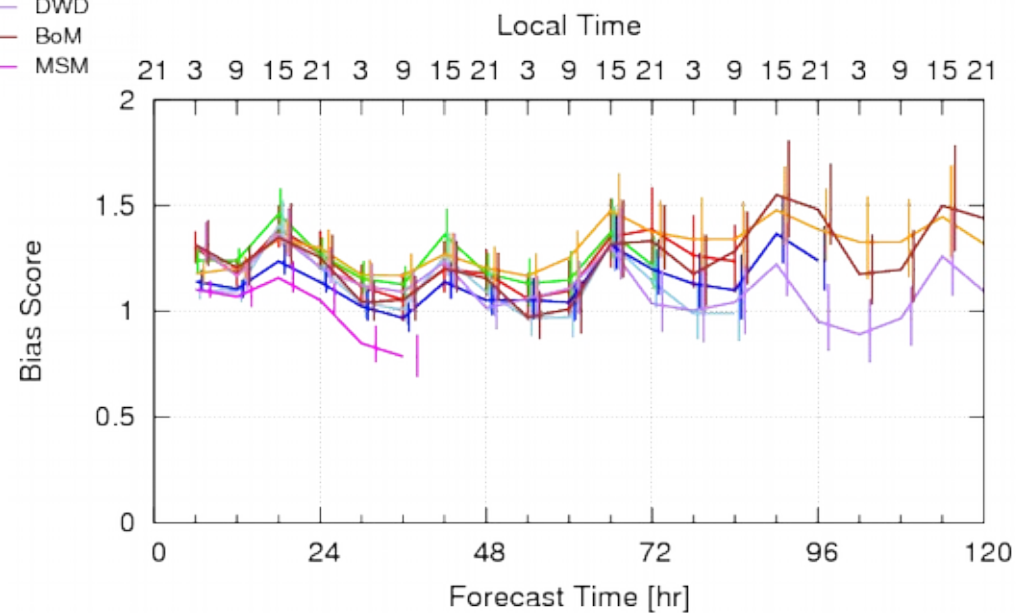
# 2015MAM: 1 mm/6hr

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

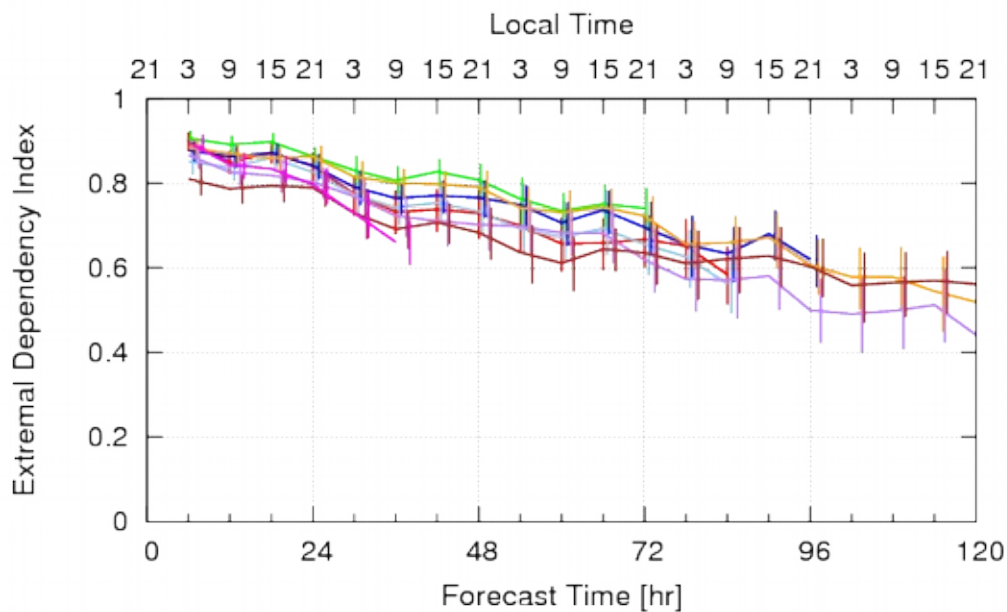
Equitable Threat Score: 1.0mm/6h 2015/03-2015/05



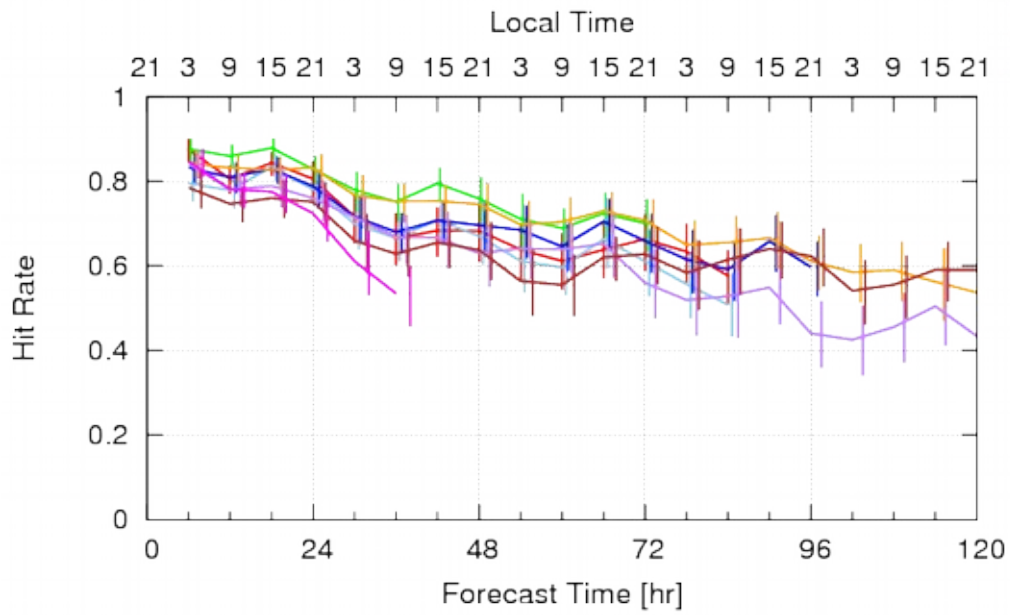
Bias Score: 1.0mm/6h 2015/03-2015/05



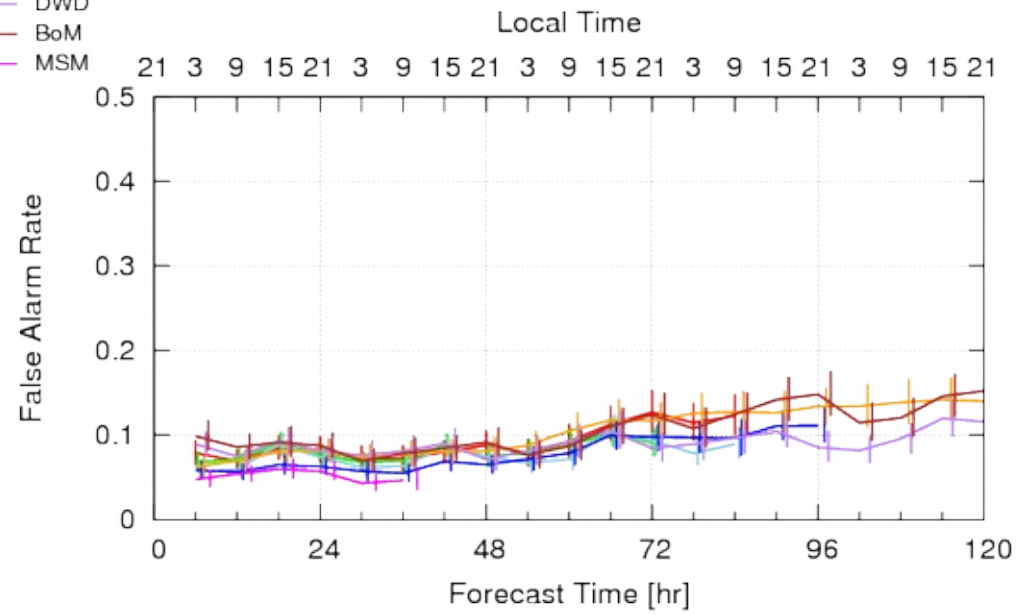
Extremal Dependency Index: 1.0mm/6h 2015/03-2015/05



Hit Rate: 1.0mm/6h 2015/03-2015/05



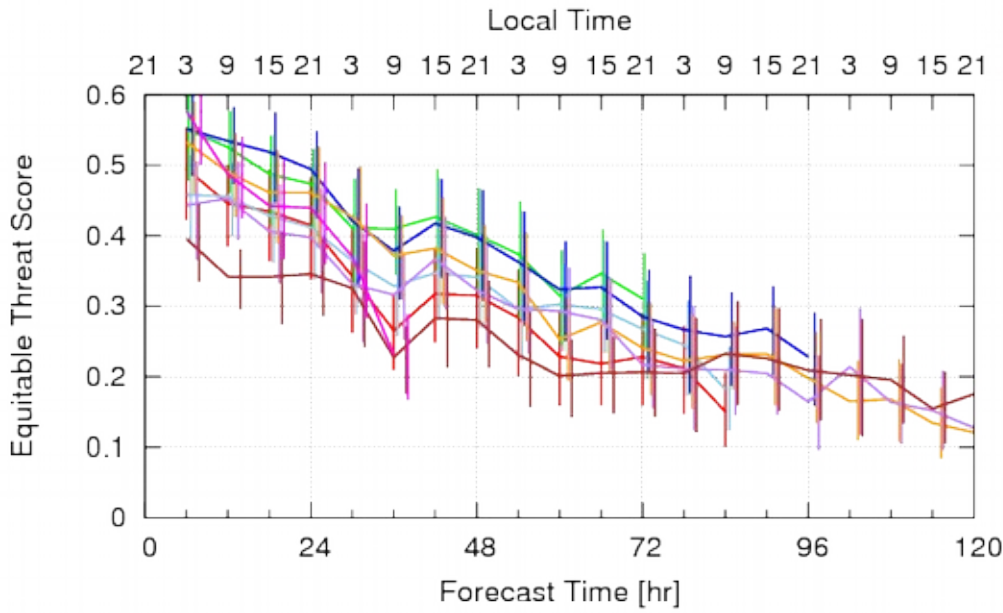
False Alarm Rate: 1.0mm/6h 2015/03-2015/05



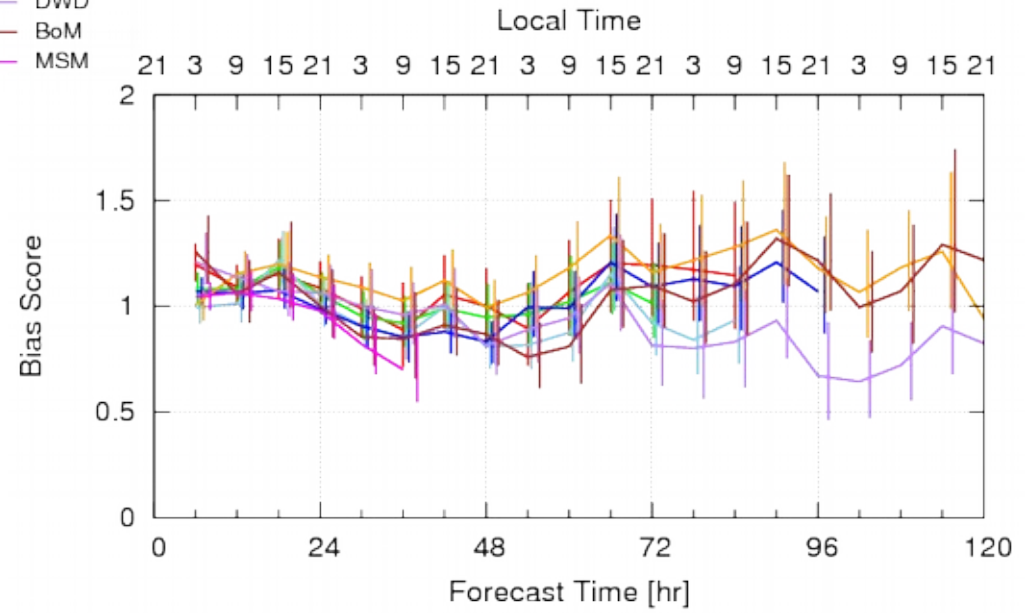
# 2015MAM: 5 mm/6hr

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

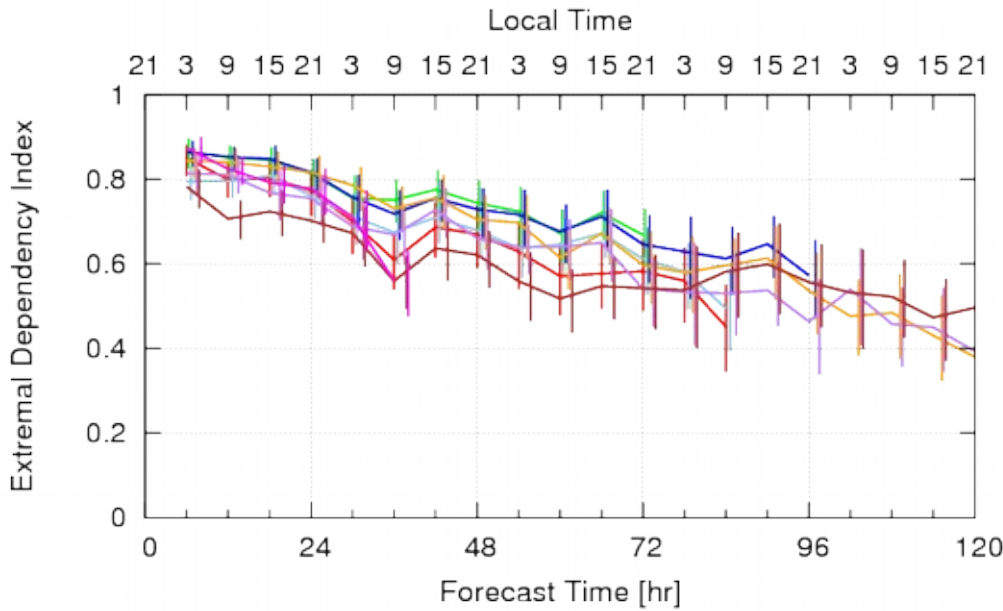
### Equitable Threat Score: 5.0mm/6h 2015/03-2015/05



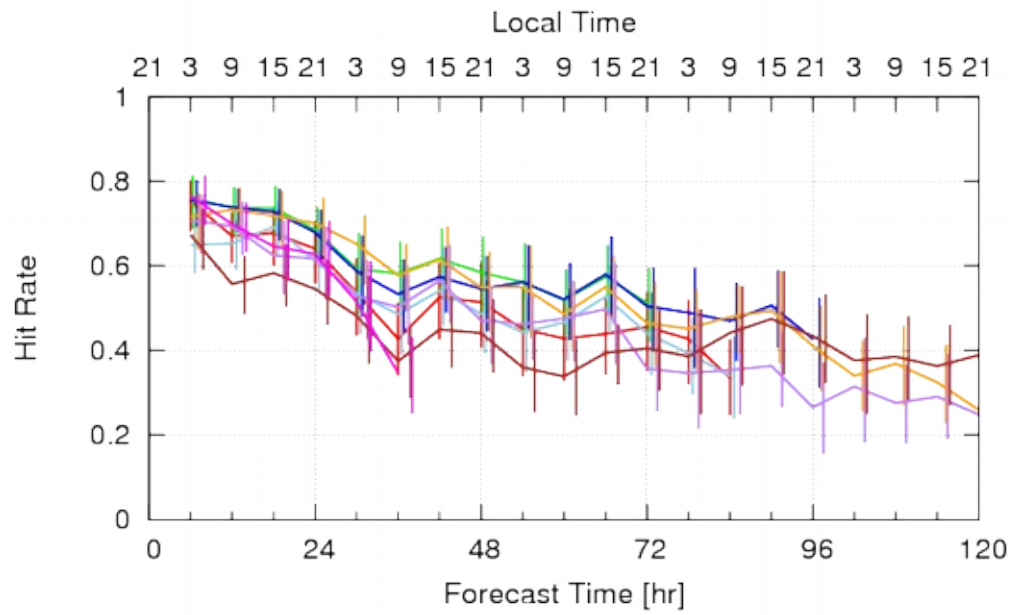
### Bias Score: 5.0mm/6h 2015/03-2015/05



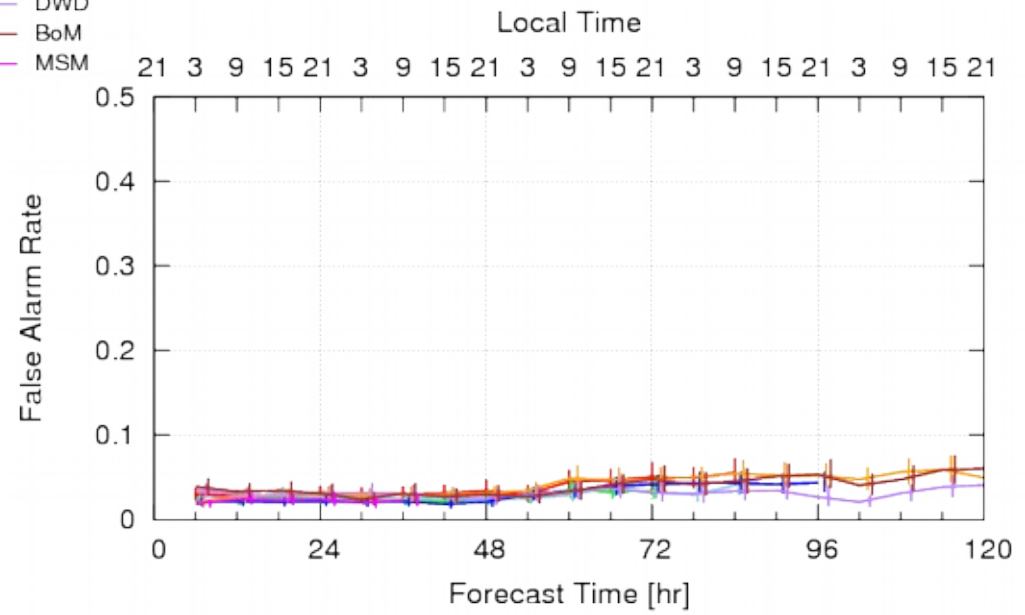
### Extremal Dependency Index: 5.0mm/6h 2015/03-2015/05



Hit Rate: 5.0mm/6h 2015/03-2015/05

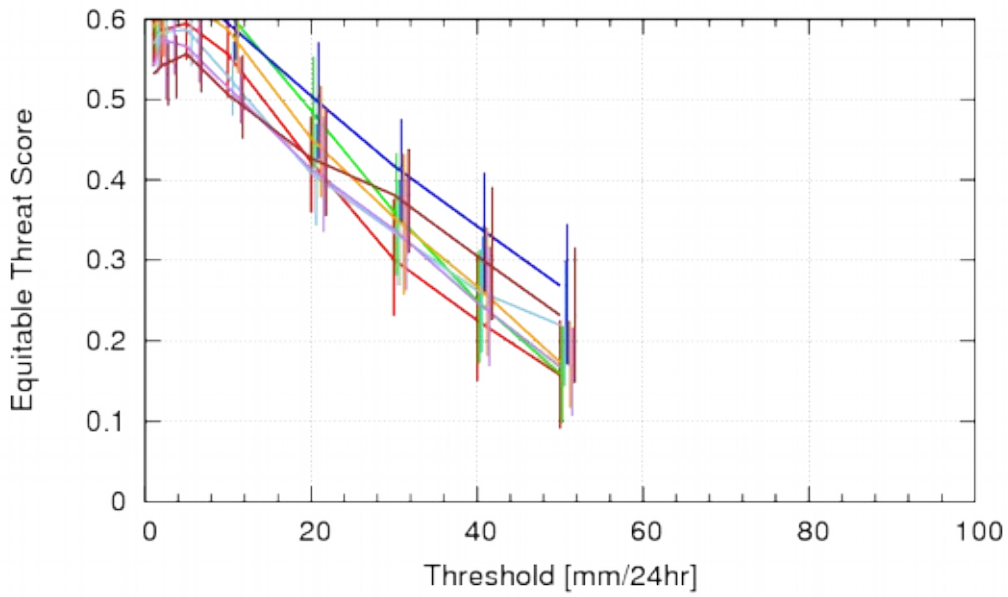


False Alarm Rate: 5.0mm/6h 2015/03-2015/05

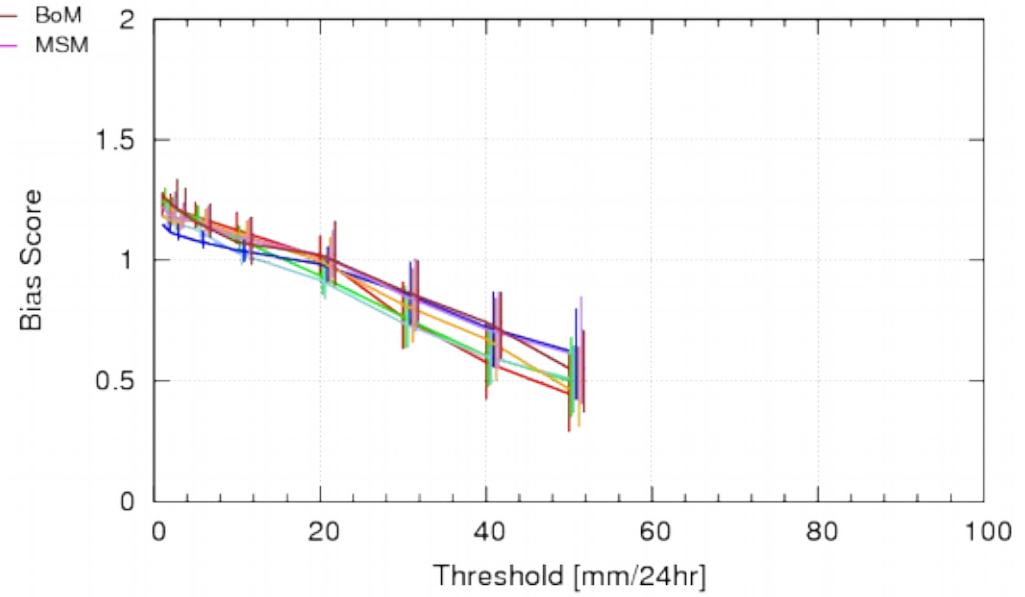


- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

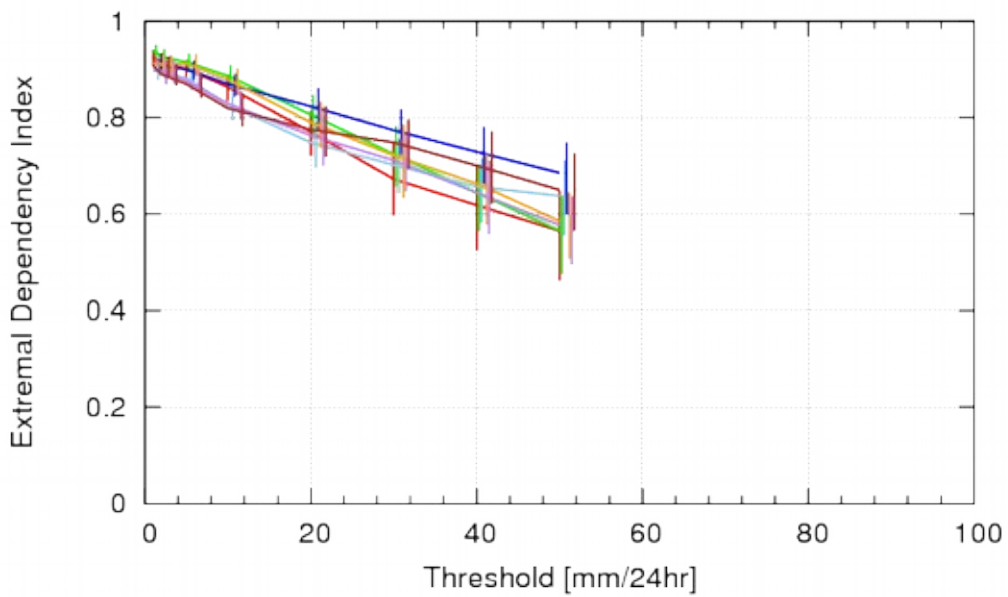
Equitable Threat Score: FT0-24 2015/03-2015/05



Bias Score: FT0-24 2015/03-2015/05



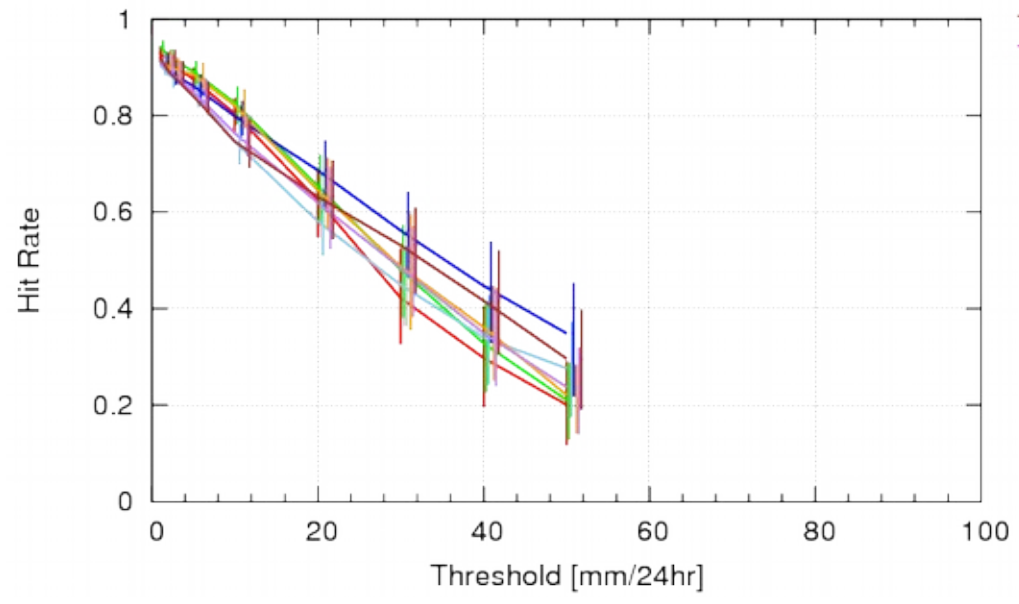
Extremal Dependency Index: FT0-24 2015/03-2015/05



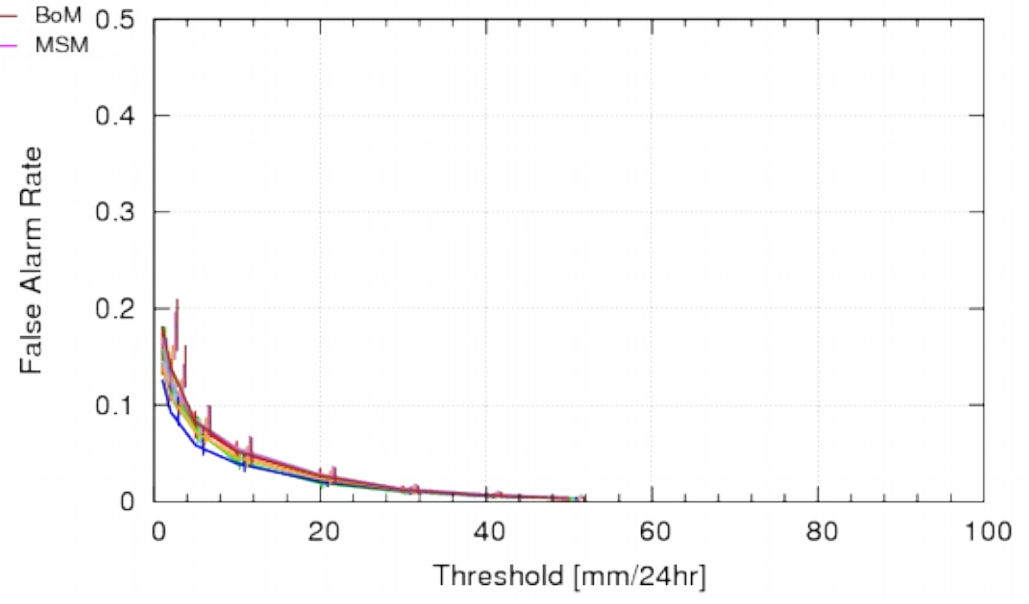


- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

Hit Rate: FT0-24 2015/03-2015/05

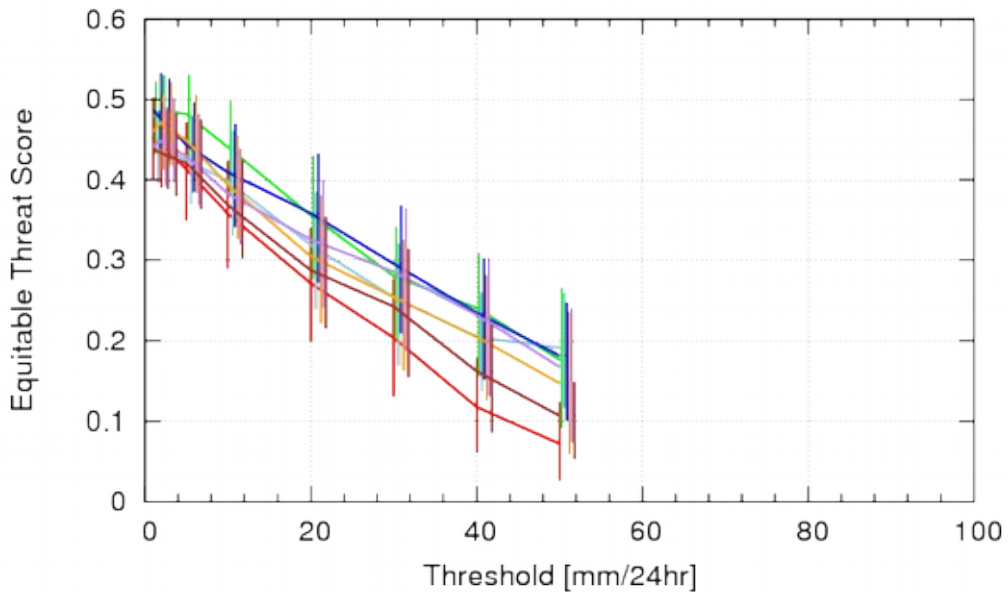


False Alarm Rate: FT0-24 2015/03-2015/05

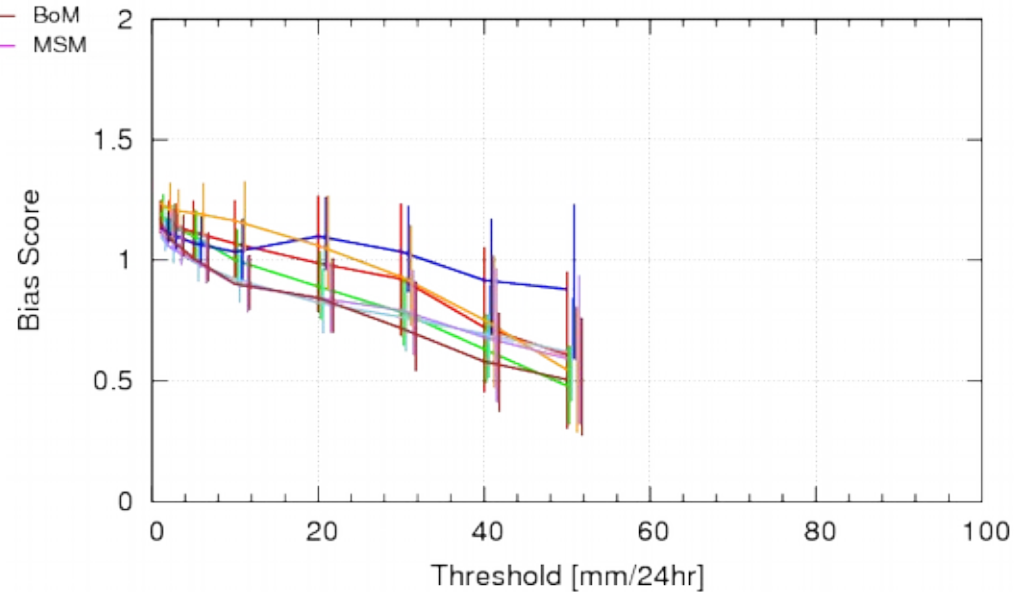


- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

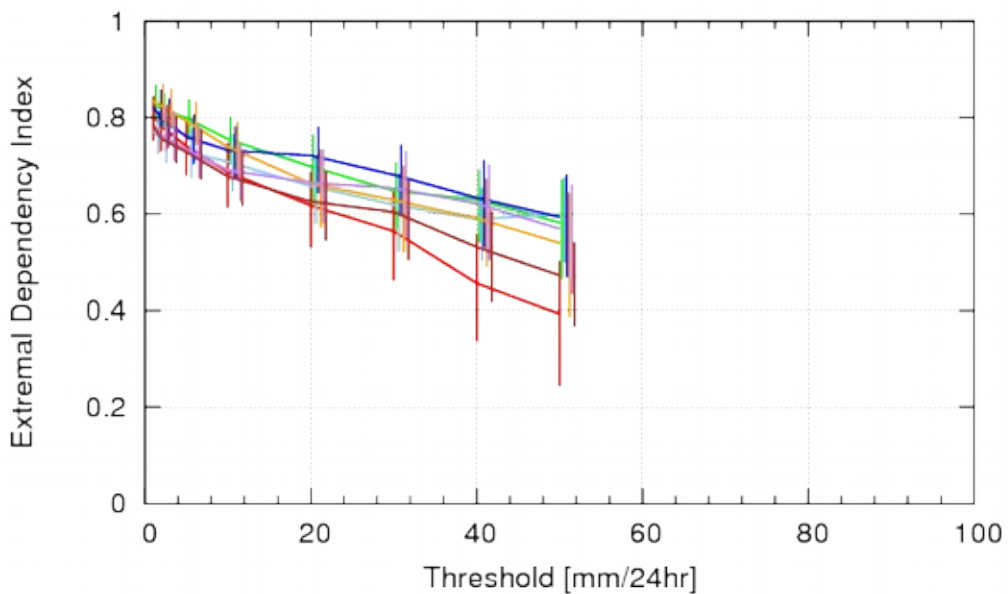
Equitable Threat Score: FT48-72 2015/03-2015/05

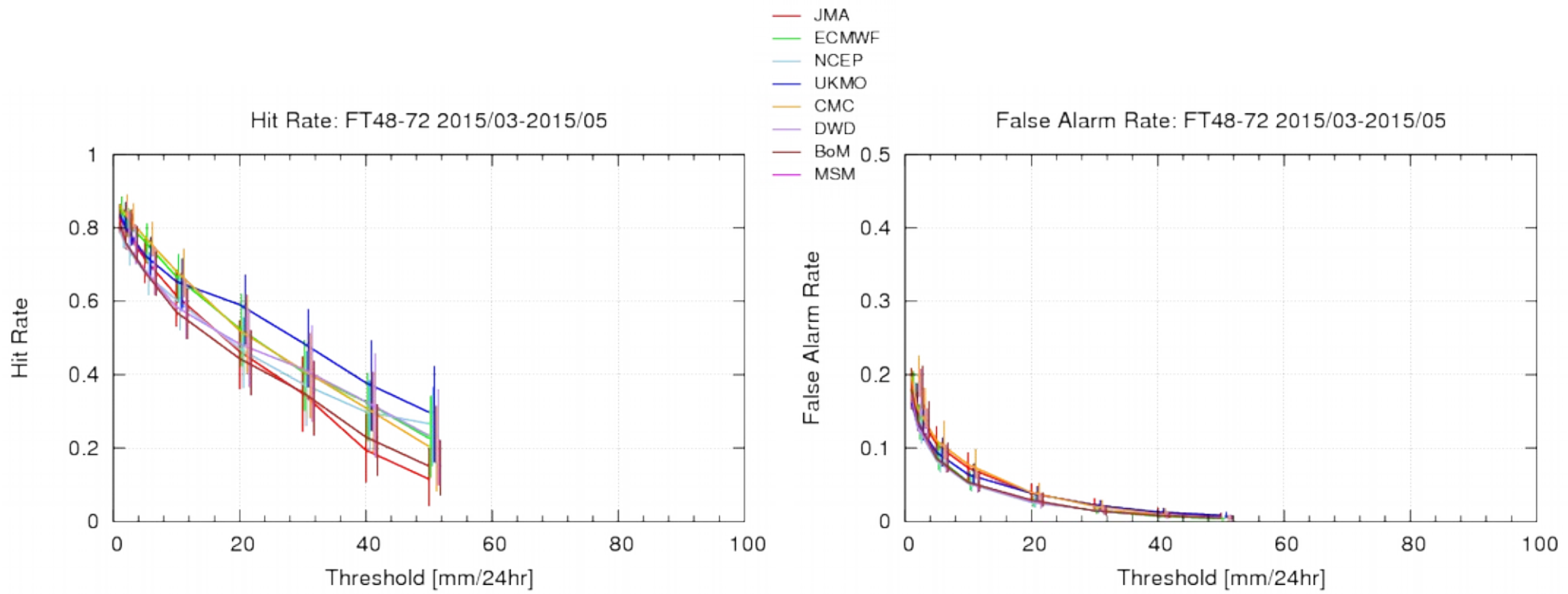


Bias Score: FT48-72 2015/03-2015/05



Extremal Dependency Index: FT48-72 2015/03-2015/05

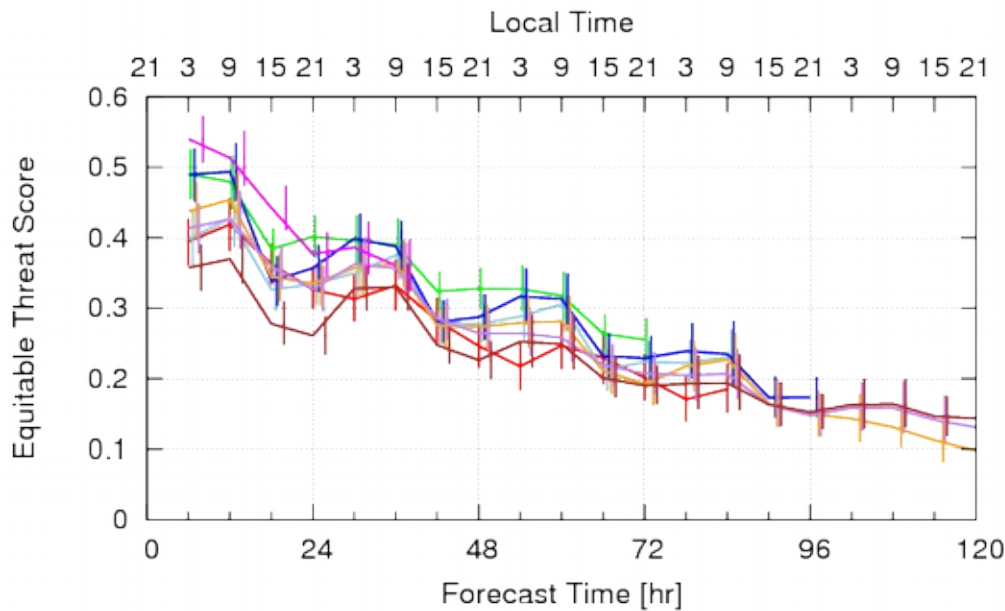




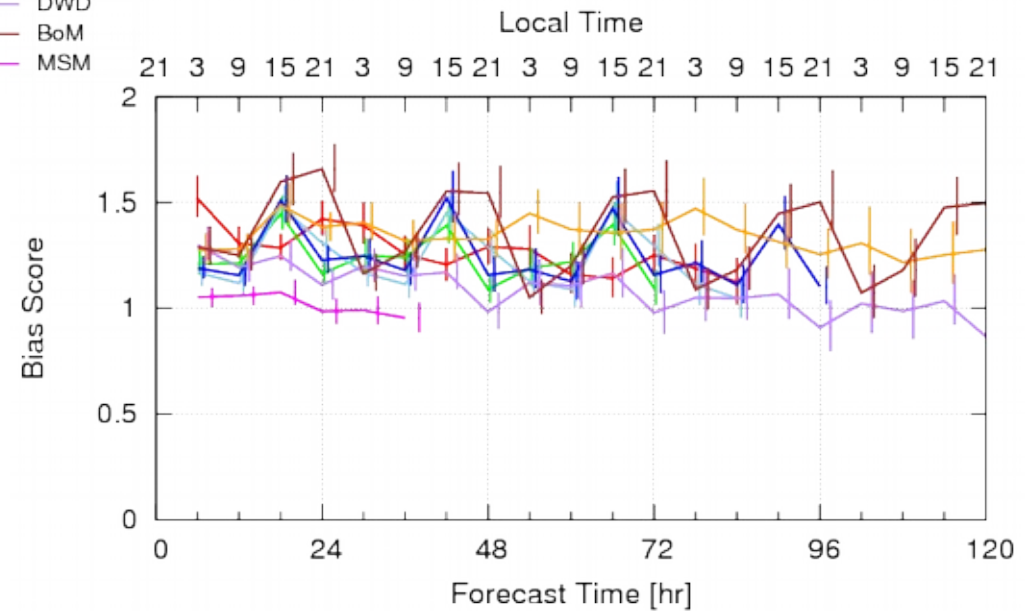
2015JJA

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

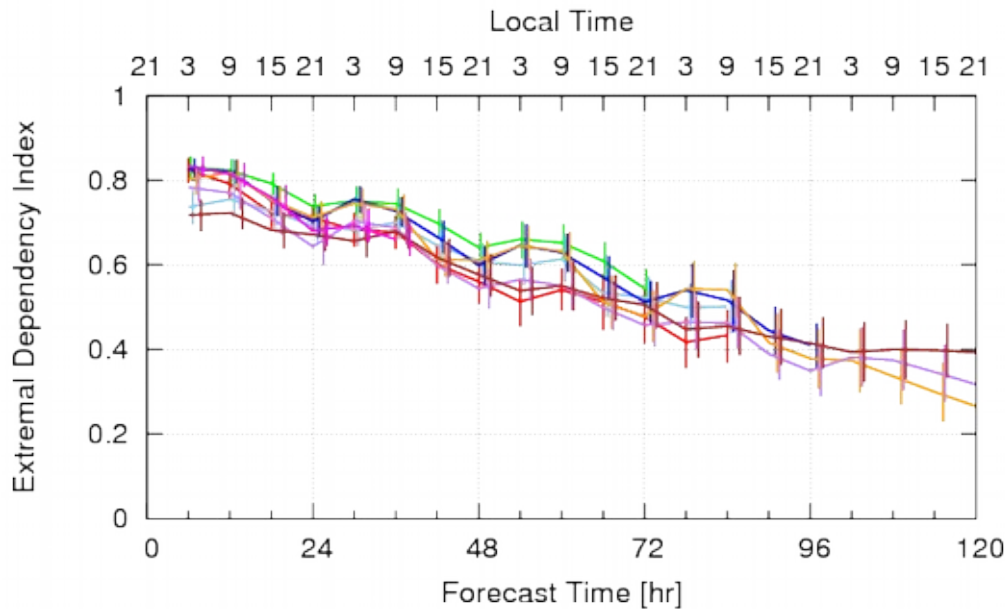
Equitable Threat Score: 1.0mm/6h 2015/06-2015/08



Bias Score: 1.0mm/6h 2015/06-2015/08

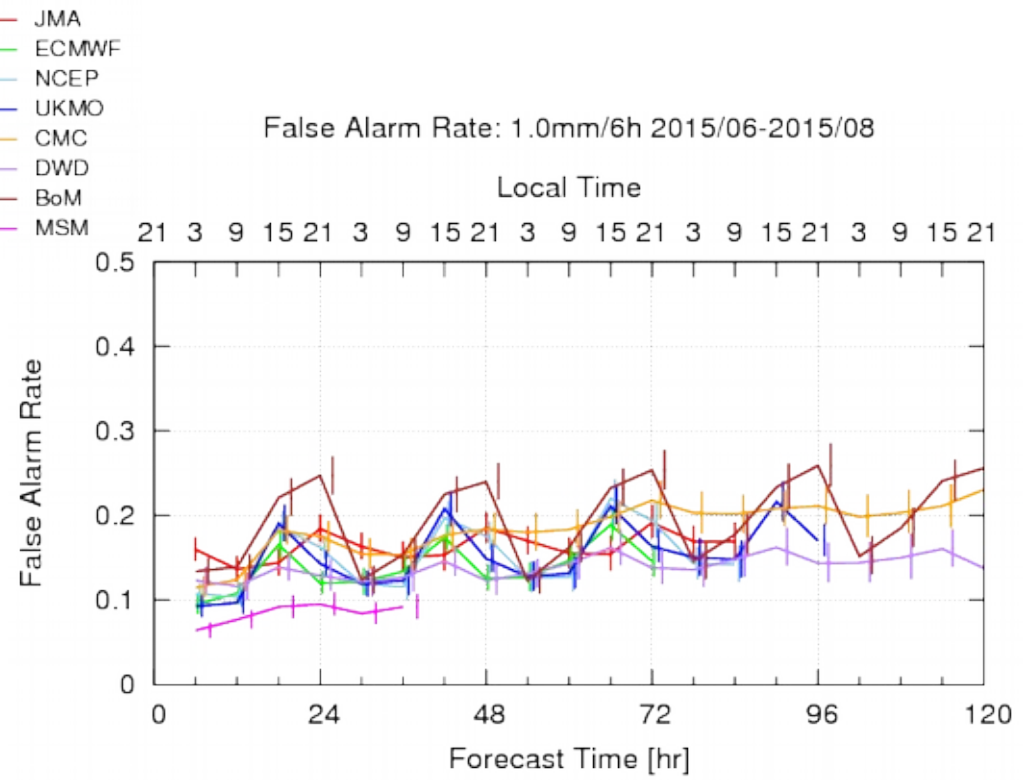
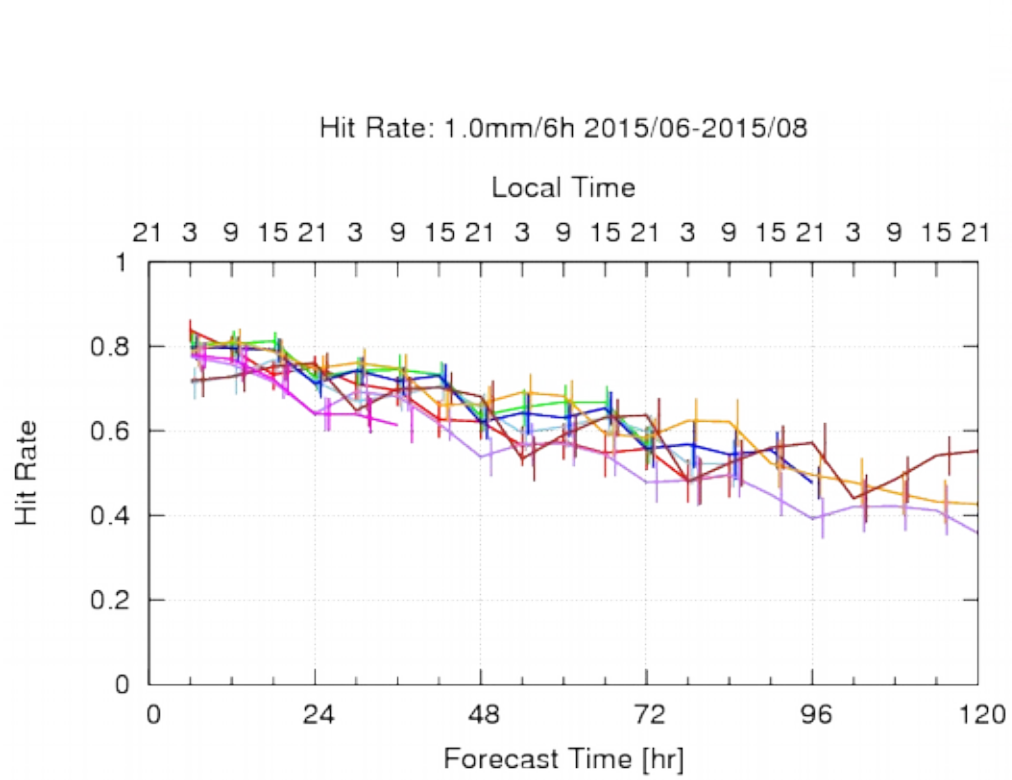


Extremal Dependency Index: 1.0mm/6h 2015/06-2015/08



- MSM performs quite better in BI.
- In BI, diurnal cycle is seen in many centers (large BI at 15 local time).
- MSM performs better in ETS on FT<24.
- ECMWF performs better in ETS and EDI.

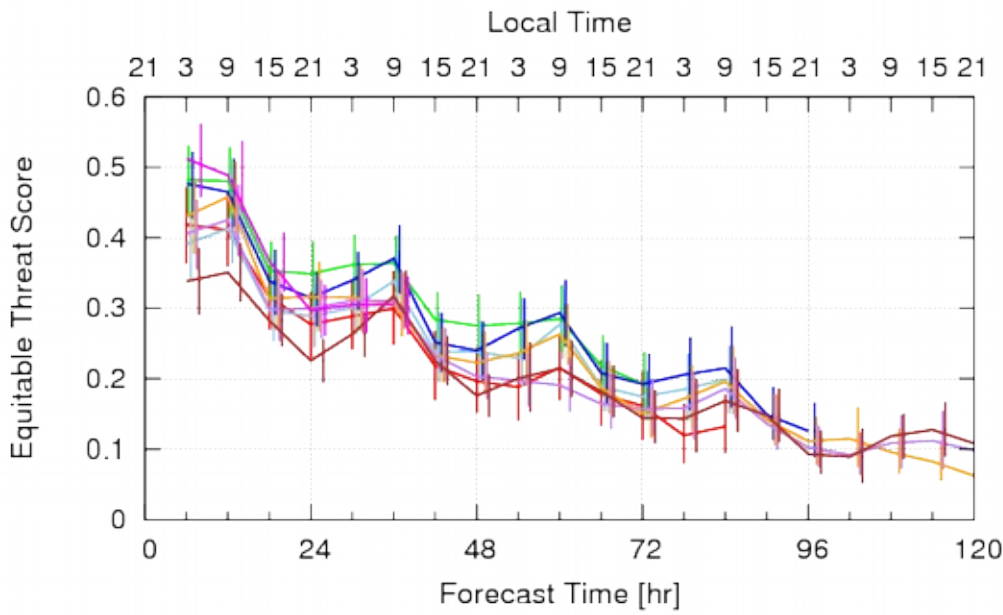




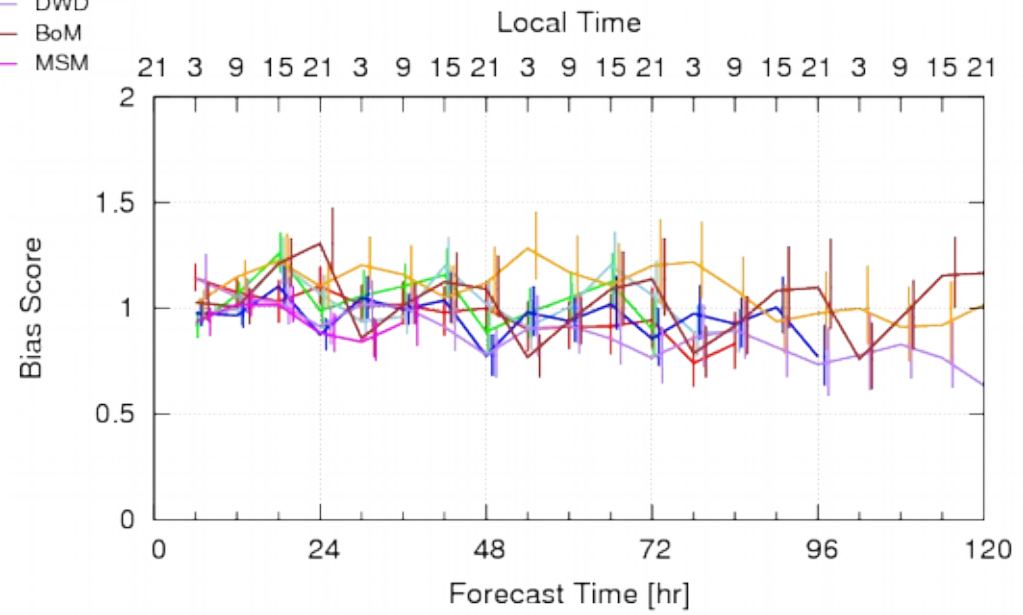
- MSM performs better in HR and FAR on  $FT < 24$ . HR of MSM decreases on  $FT \geq 24$ , which may cause descent of ETS and EDI. JMA, which provides the boundary condition of MSM, does not show such a behavior.

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

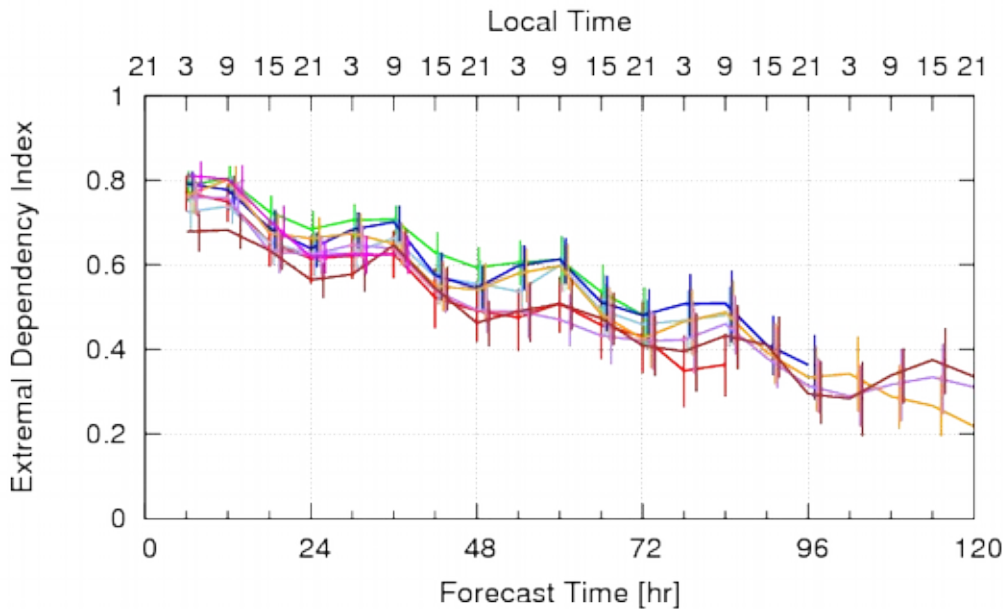
Equitable Threat Score: 5.0mm/6h 2015/06-2015/08



Bias Score: 5.0mm/6h 2015/06-2015/08

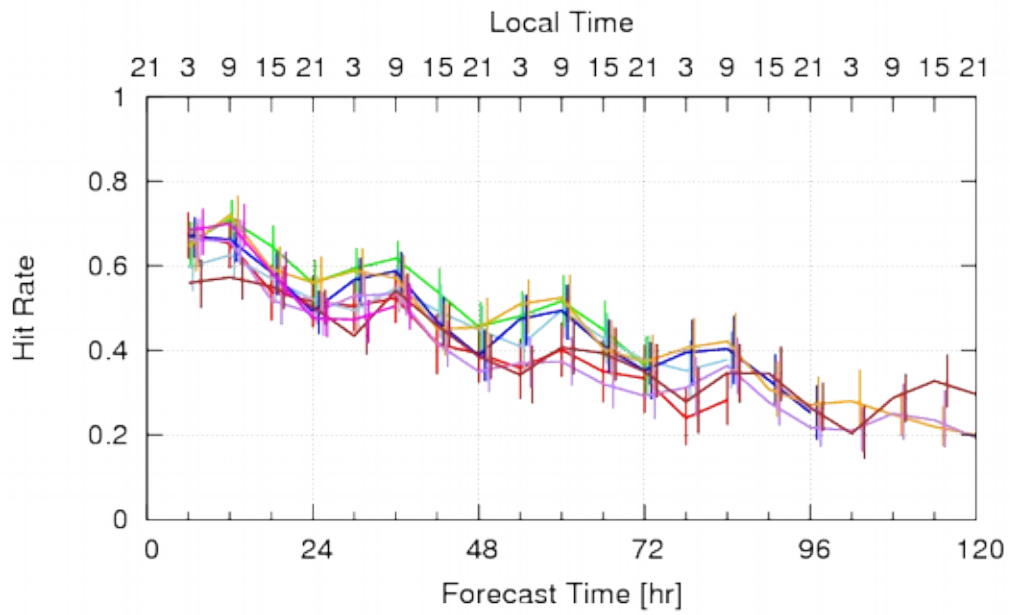


Extremal Dependency Index: 5.0mm/6h 2015/06-2015/08

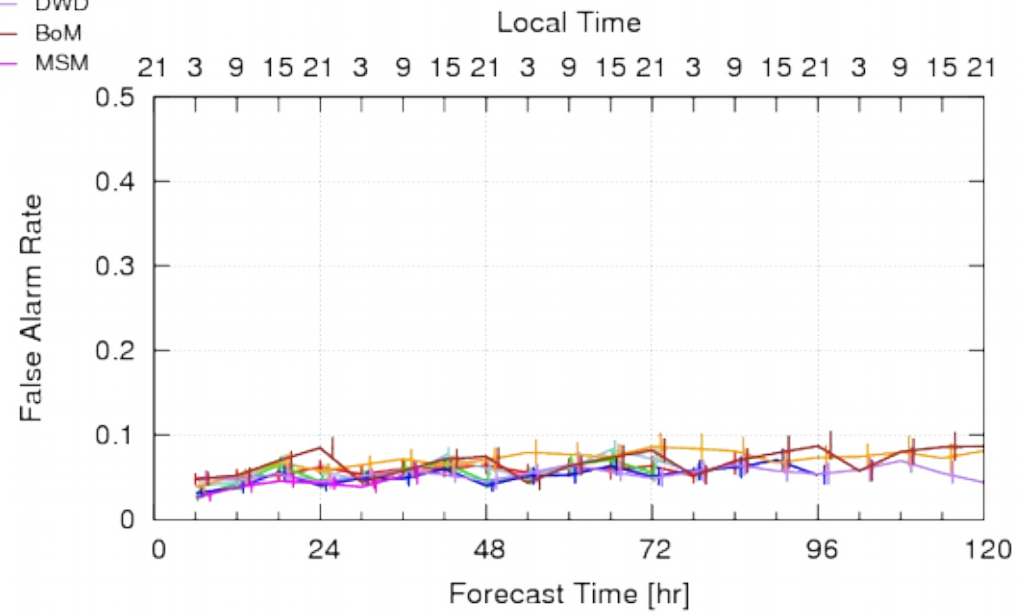


○ECMWF performs better in ETS and EDI.

Hit Rate: 5.0mm/6h 2015/06-2015/08

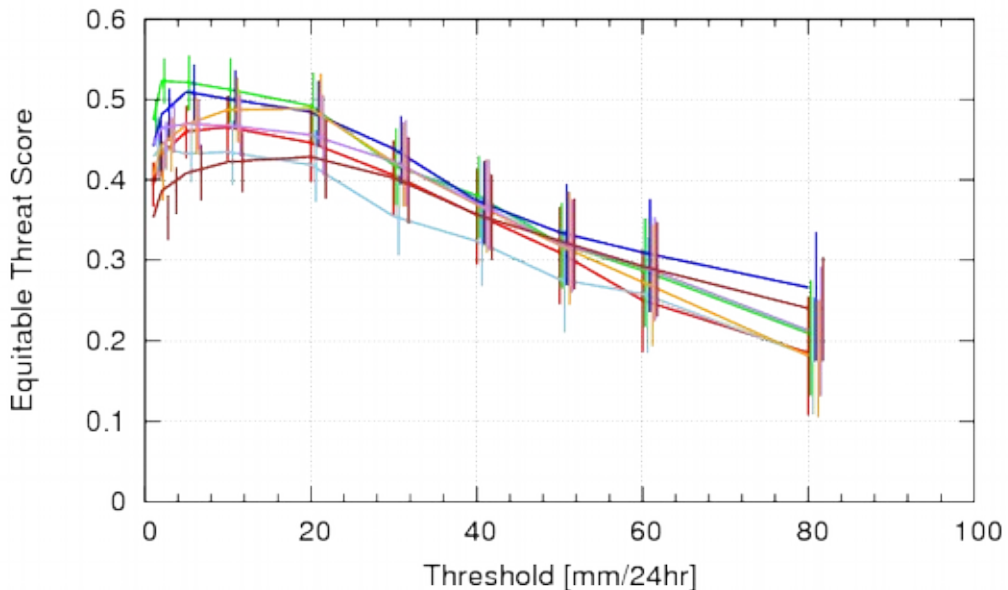


False Alarm Rate: 5.0mm/6h 2015/06-2015/08

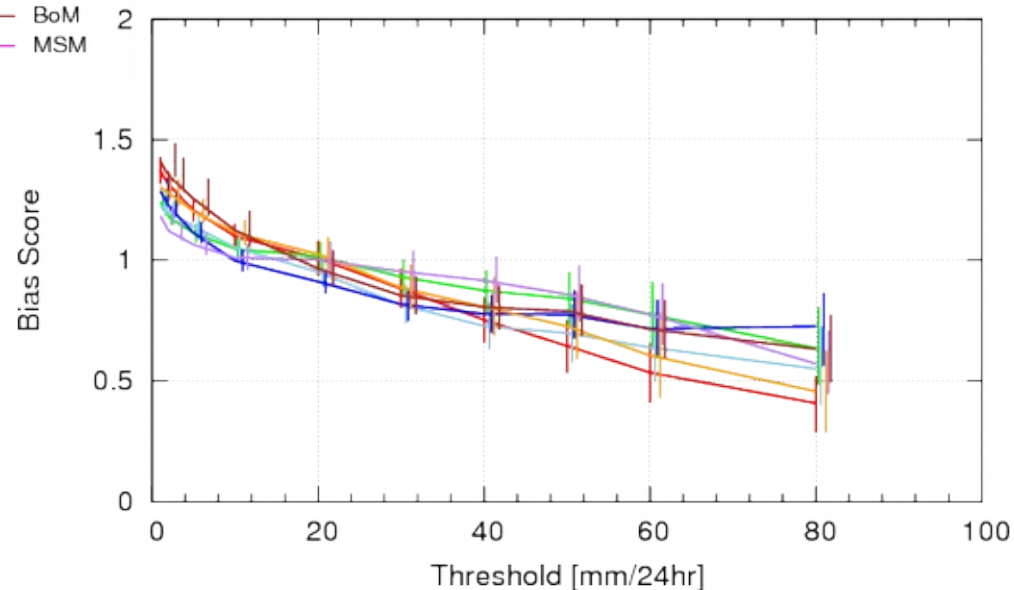


- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

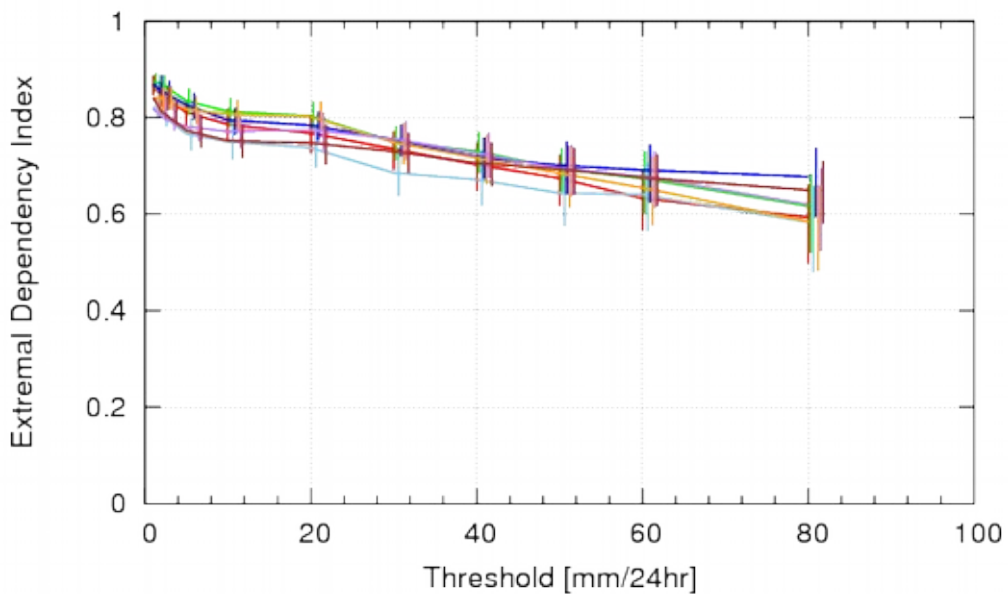
Equitable Threat Score: FT0-24 2015/06-2015/08



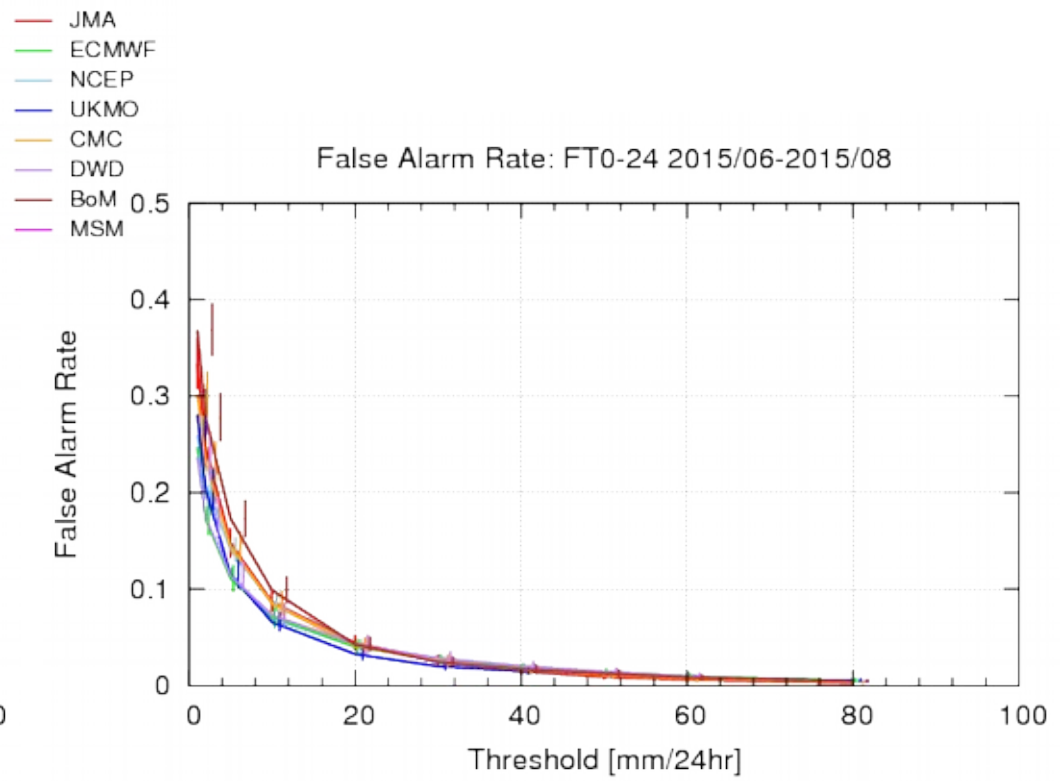
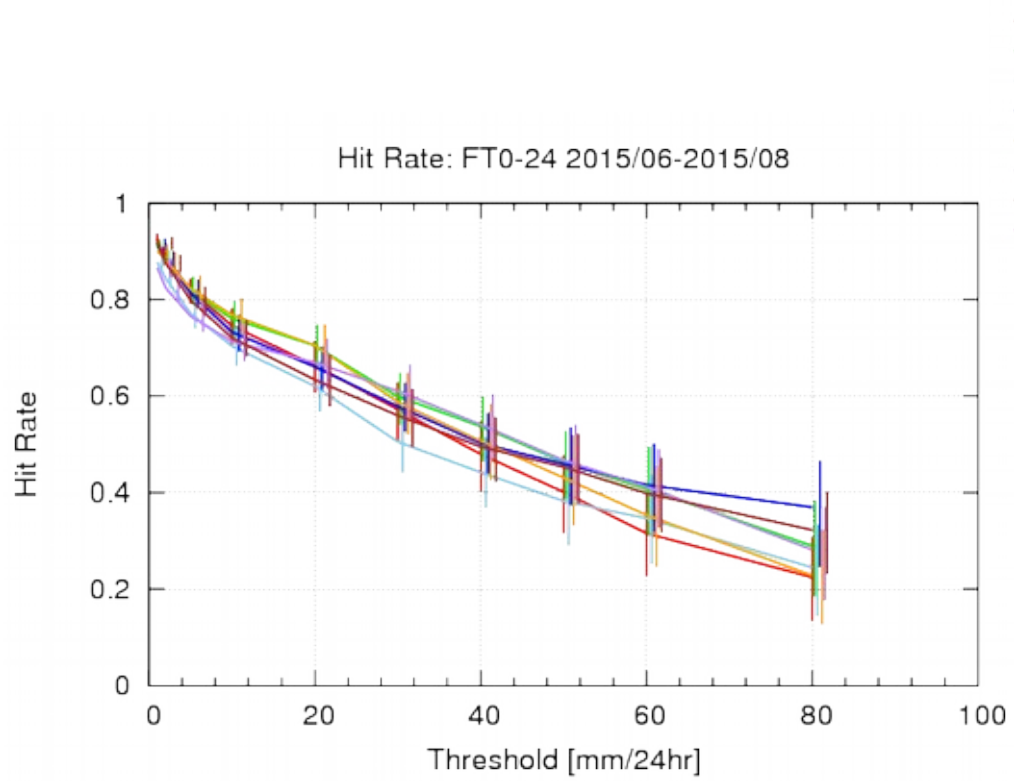
Bias Score: FT0-24 2015/06-2015/08



Extremal Dependency Index: FT0-24 2015/06-2015/08



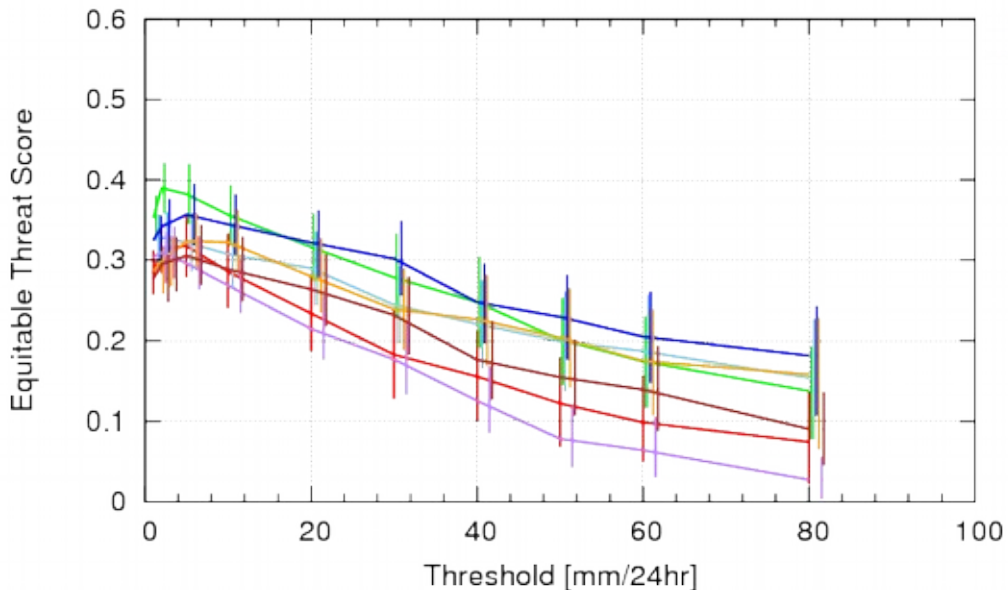
- ECMWF performs better in ETS and EDI on  $\leq 50$  mm/24hr.
- UKMO performs better in ETS and EDI.



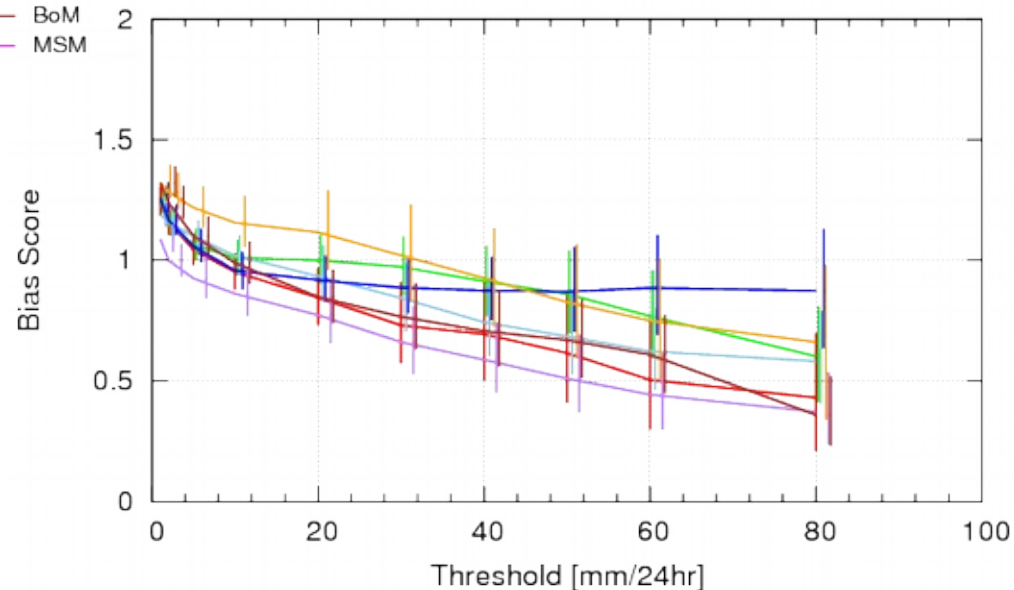


- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

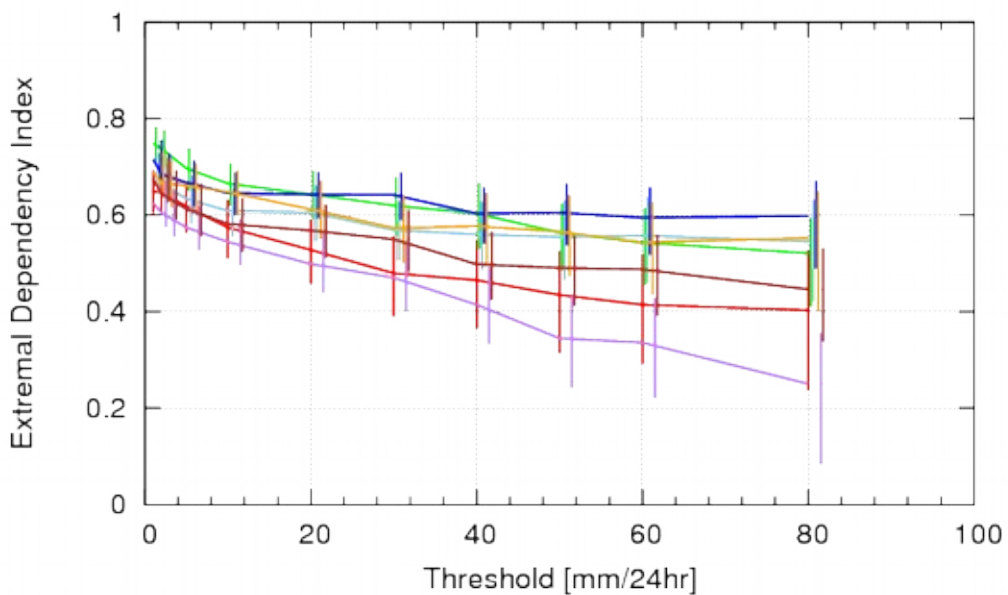
Equitable Threat Score: FT48-72 2015/06-2015/08



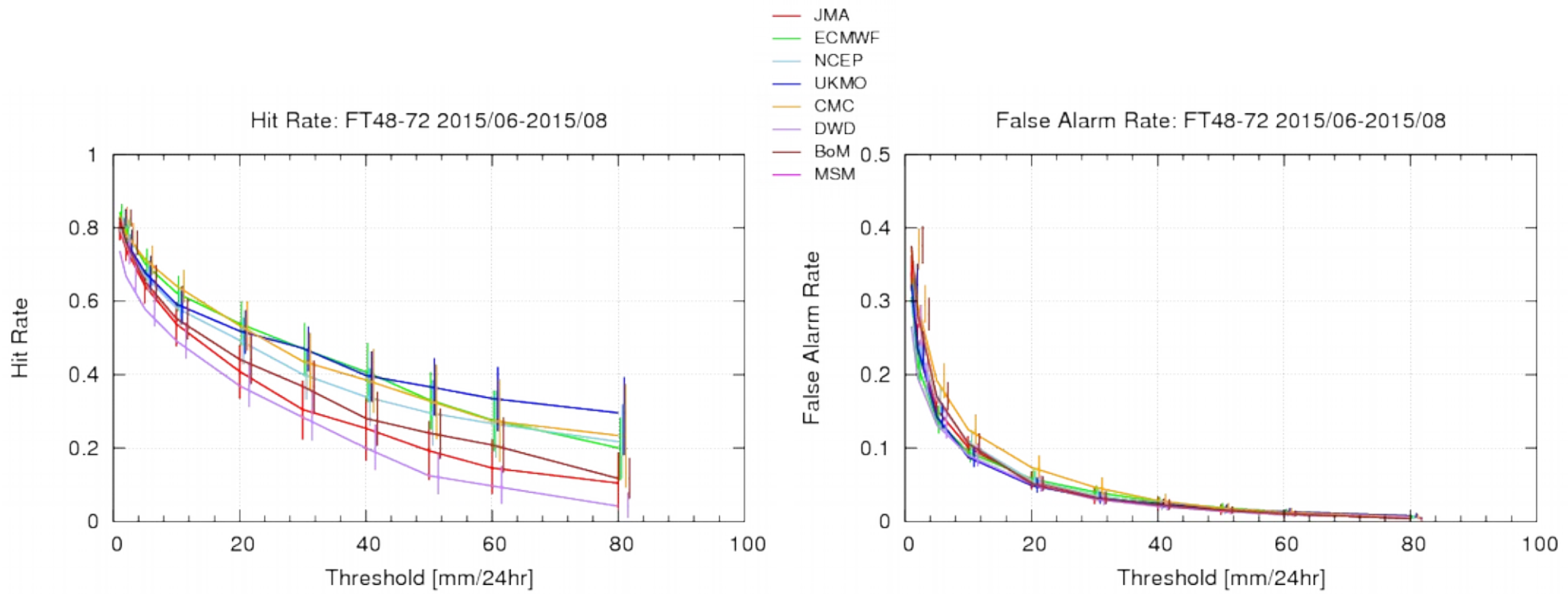
Bias Score: FT48-72 2015/06-2015/08



Extremal Dependency Index: FT48-72 2015/06-2015/08



○ECMWF performs better in ETS and EDI on  $\leq 50$  mm/24hr.  
 ○UKMO performs better in ETS and EDI.

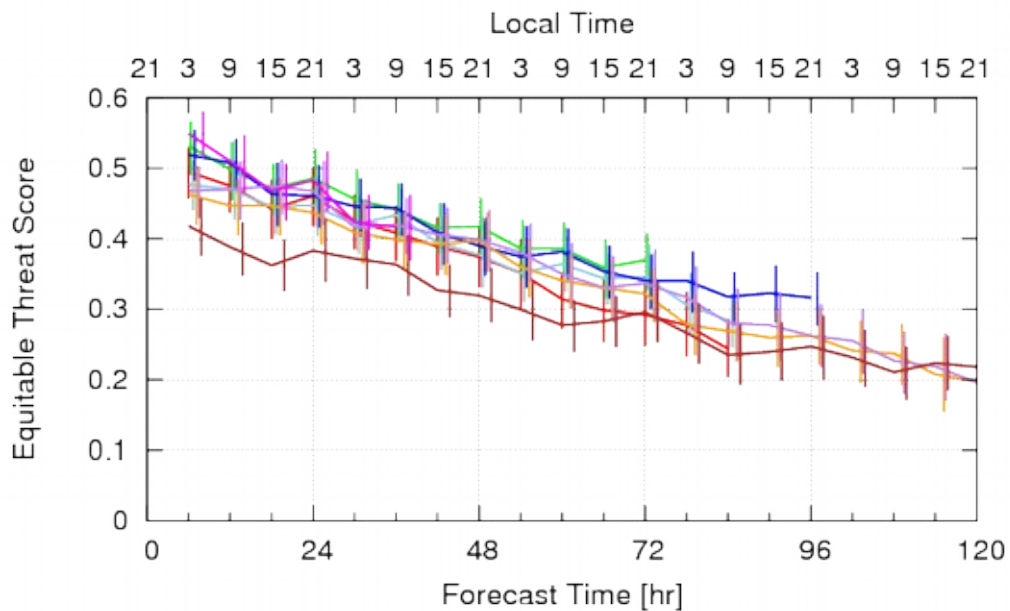


2015SON

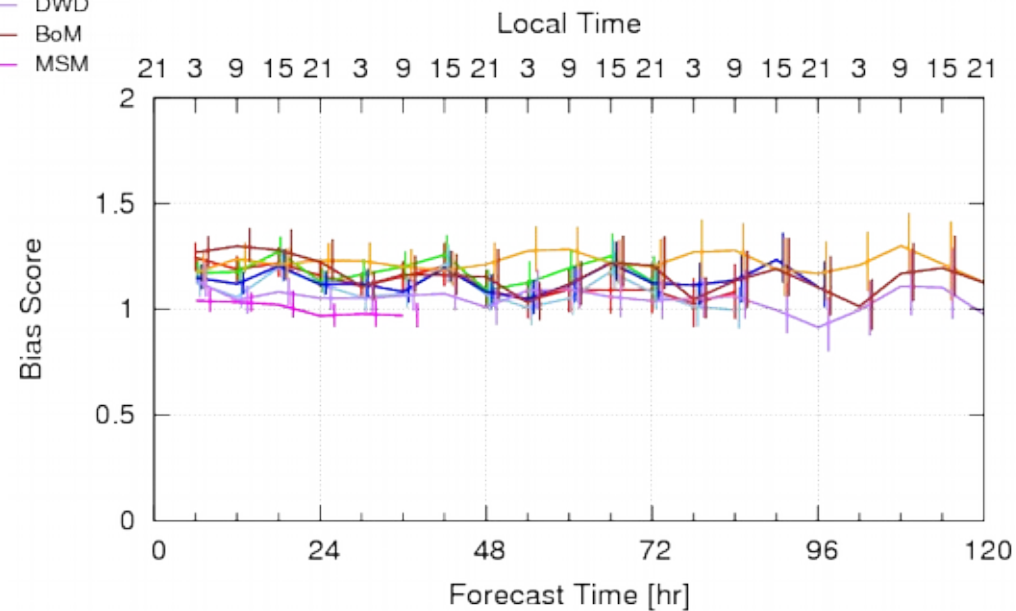
# 2015SON: 1 mm/6hr

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

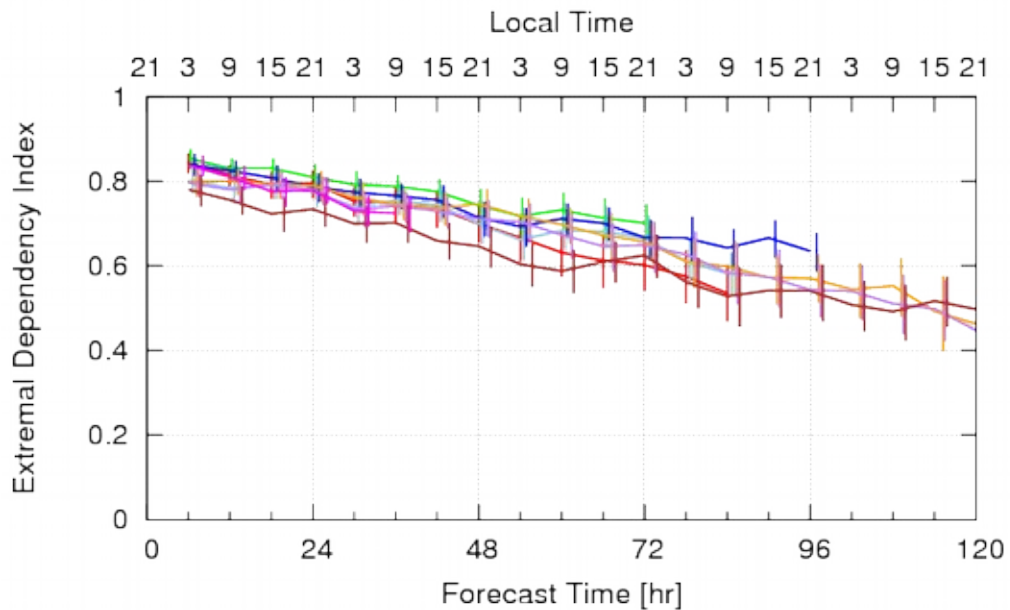
### Equitable Threat Score: 1.0mm/6h 2015/09-2015/11



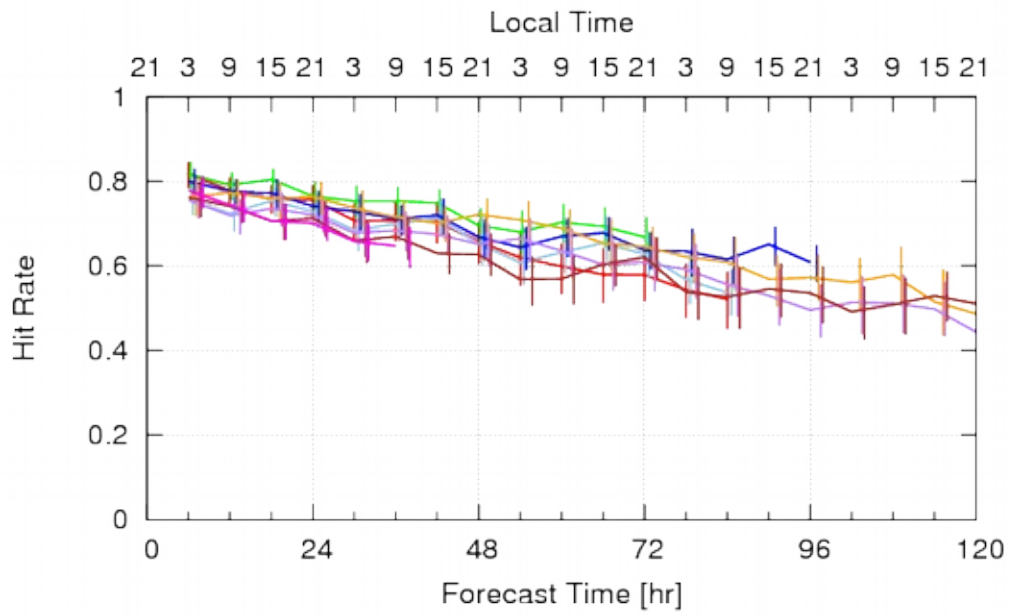
### Bias Score: 1.0mm/6h 2015/09-2015/11



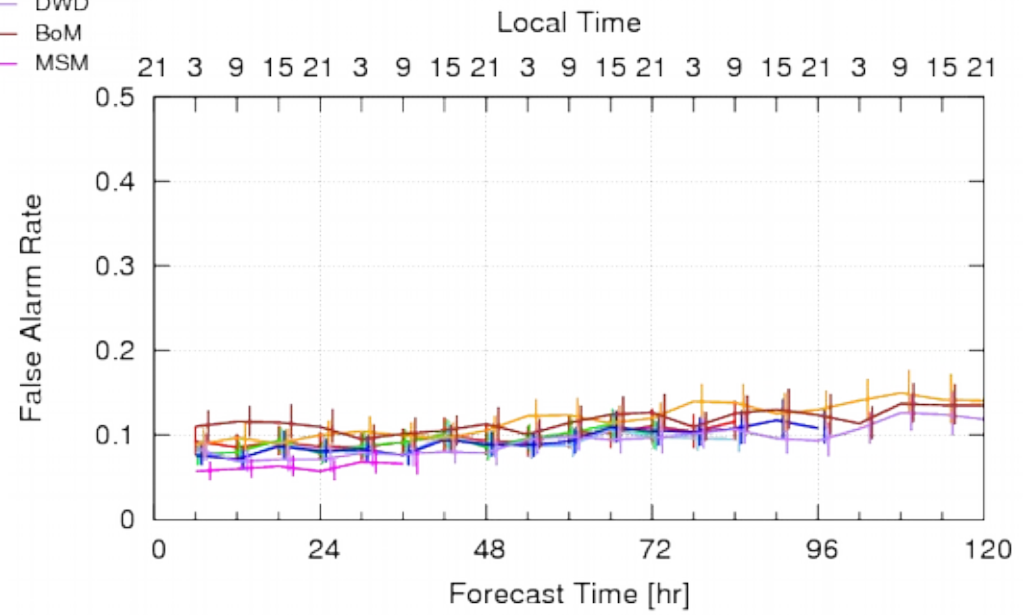
### Extremal Dependency Index: 1.0mm/6h 2015/09-2015/11



Hit Rate: 1.0mm/6h 2015/09-2015/11



False Alarm Rate: 1.0mm/6h 2015/09-2015/11

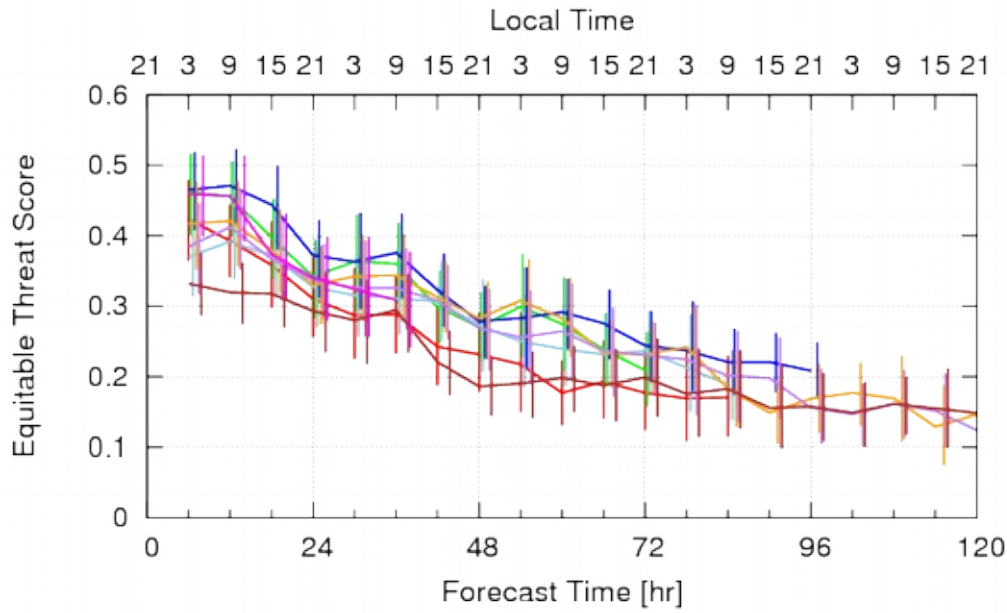




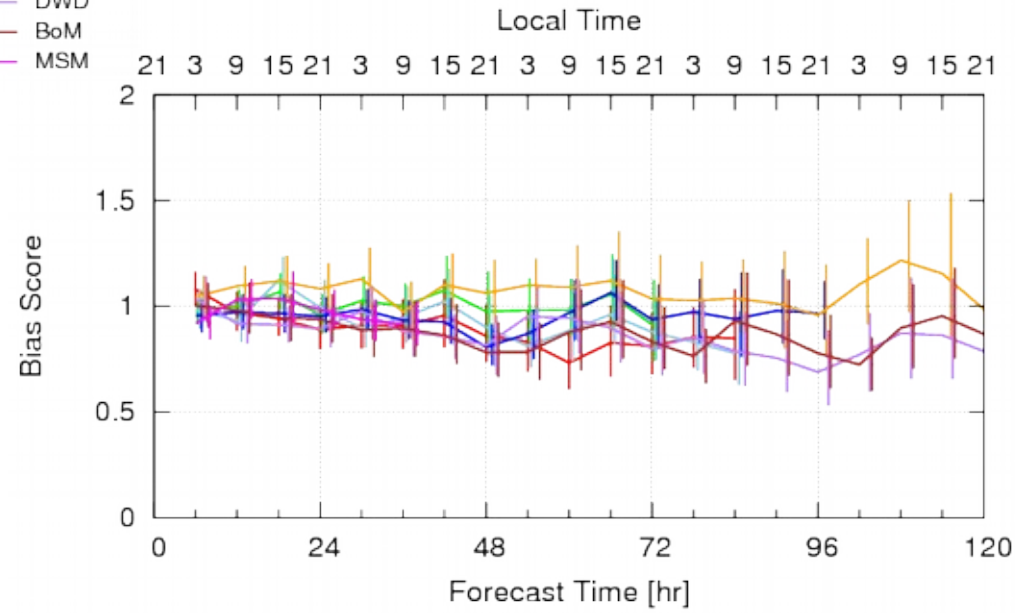
# 2015SON: 5 mm/6hr

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

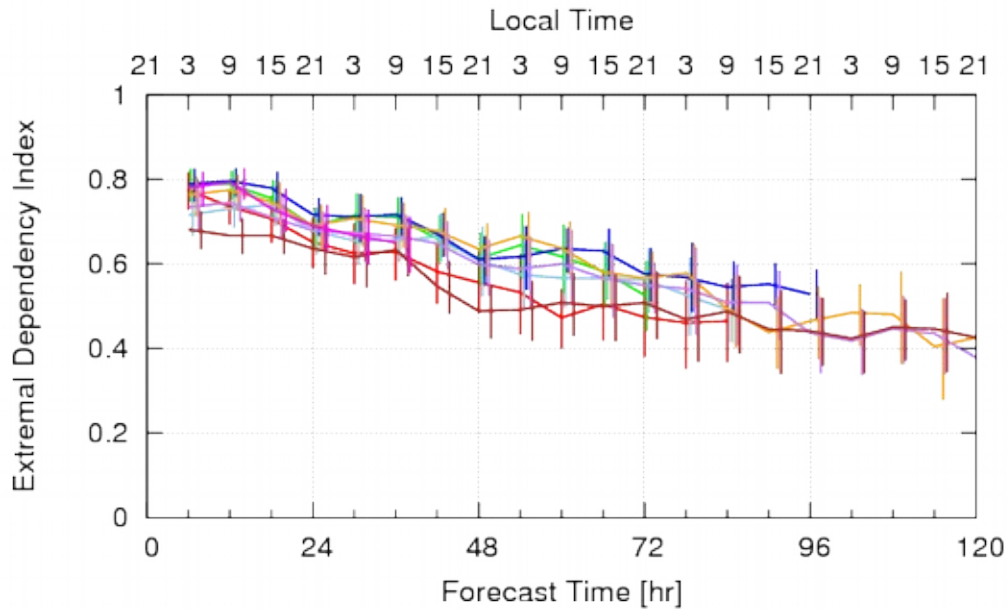
Equitable Threat Score: 5.0mm/6h 2015/09-2015/11



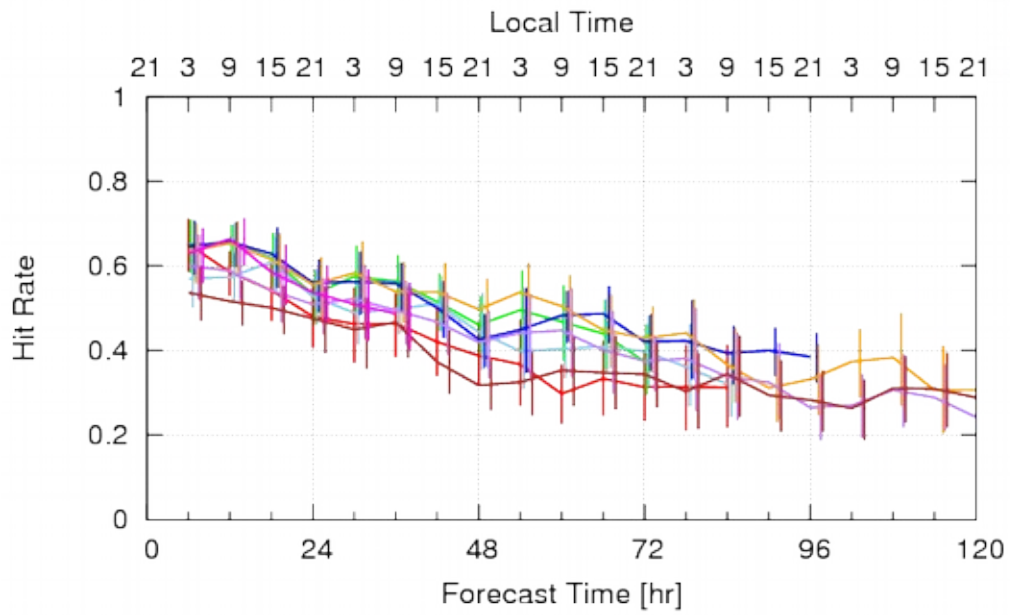
Bias Score: 5.0mm/6h 2015/09-2015/11



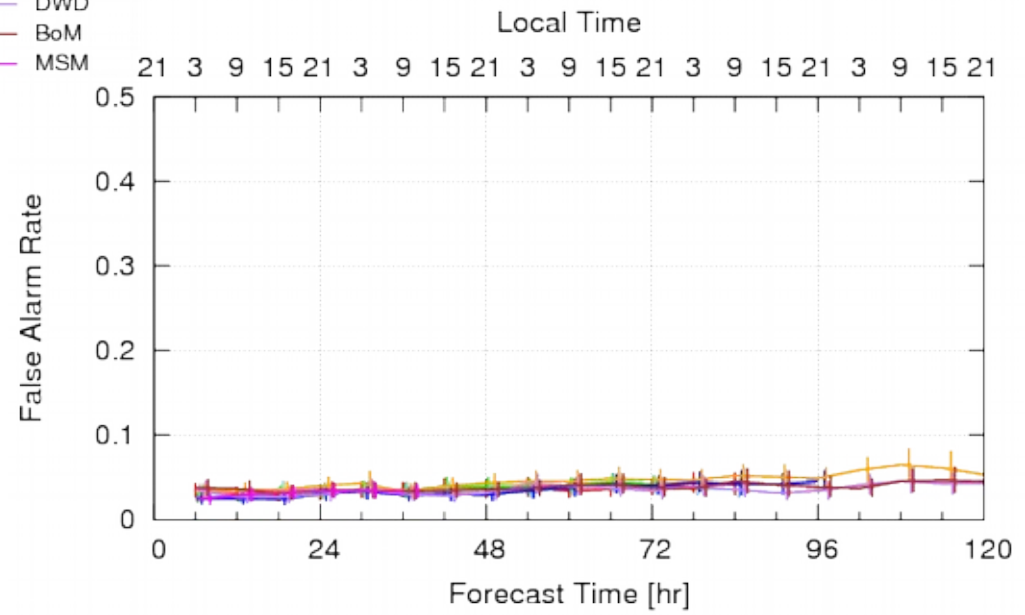
Extremal Dependency Index: 5.0mm/6h 2015/09-2015/11



Hit Rate: 5.0mm/6h 2015/09-2015/11



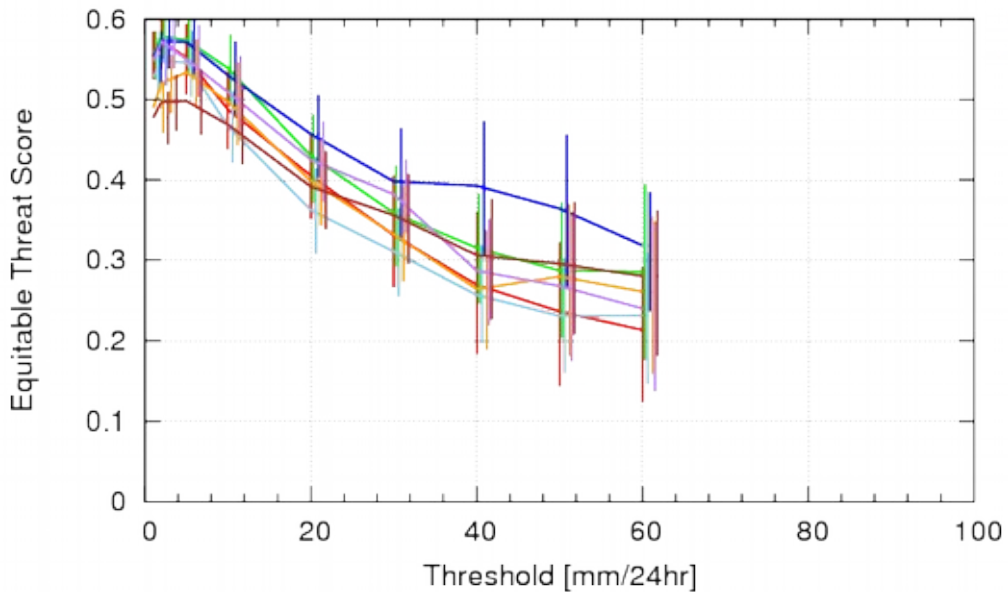
False Alarm Rate: 5.0mm/6h 2015/09-2015/11



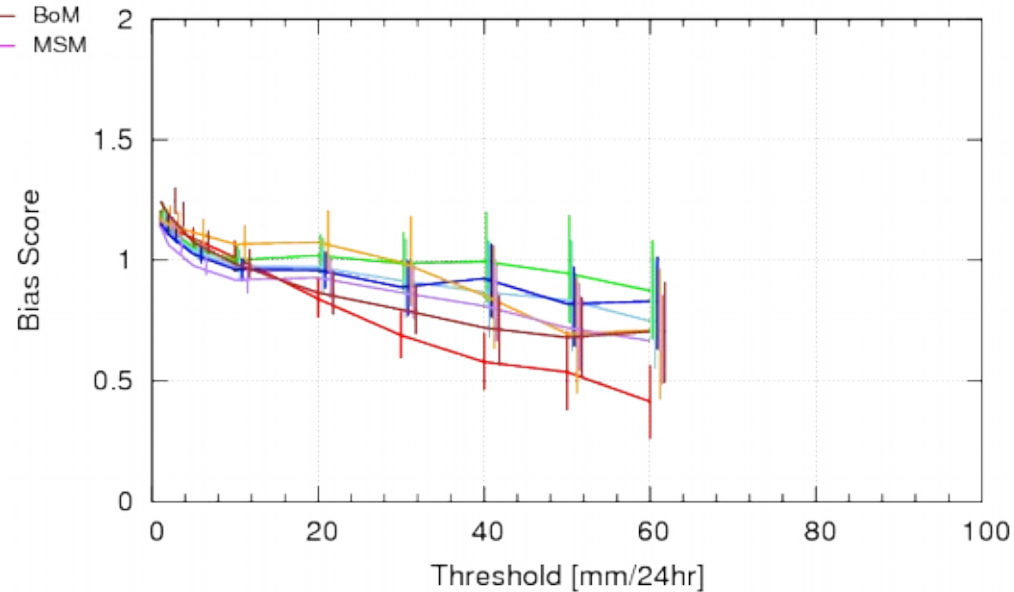
- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

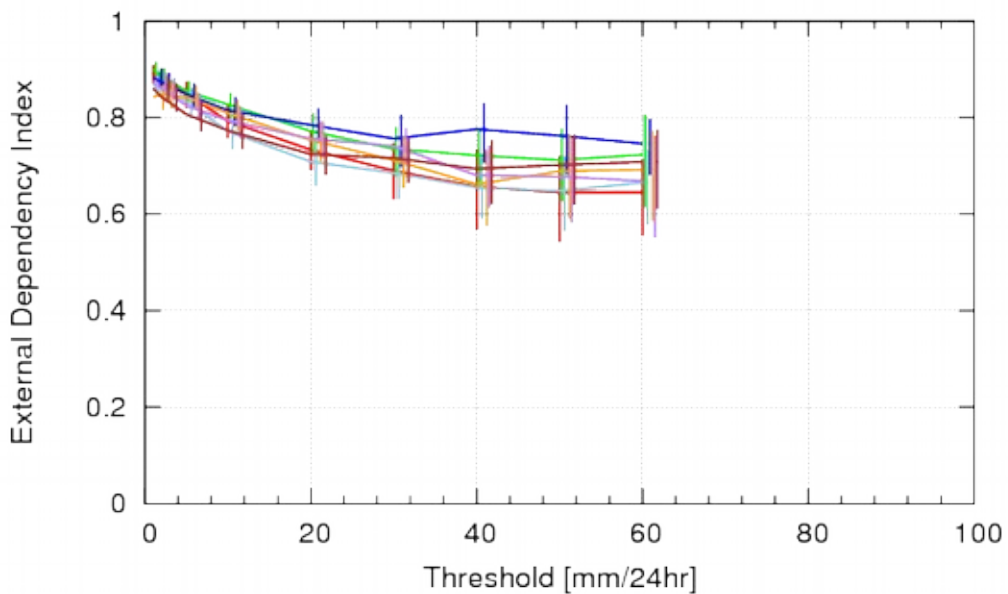
Equitable Threat Score: FT0-24 2015/09-2015/11

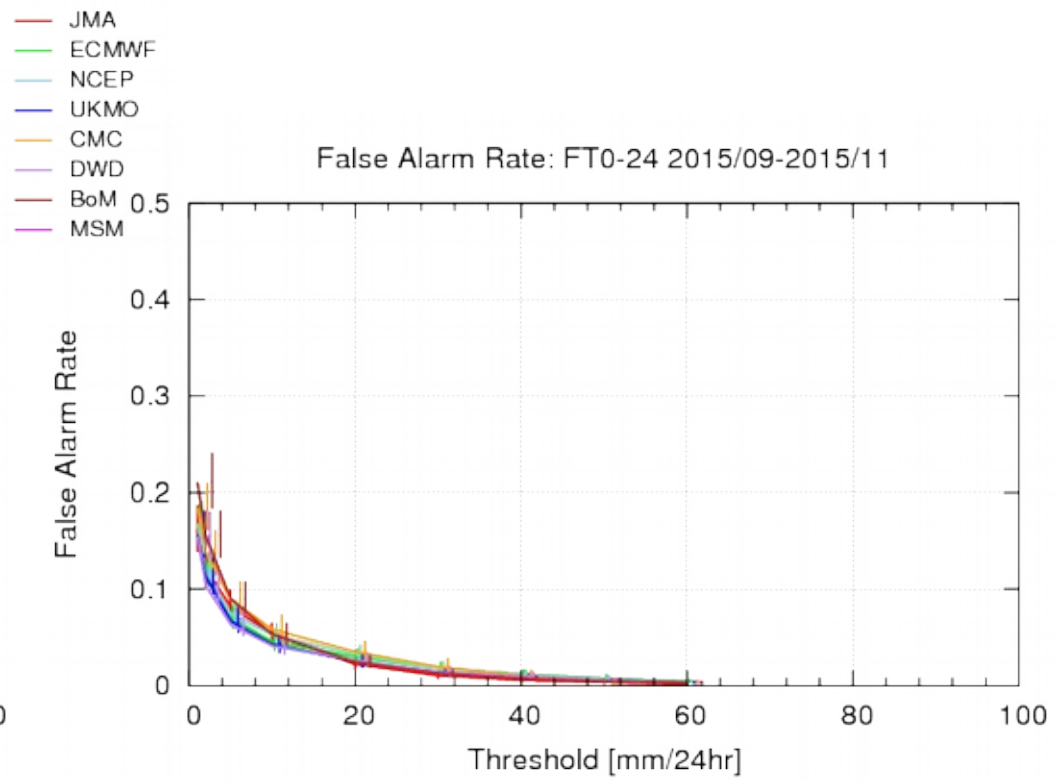
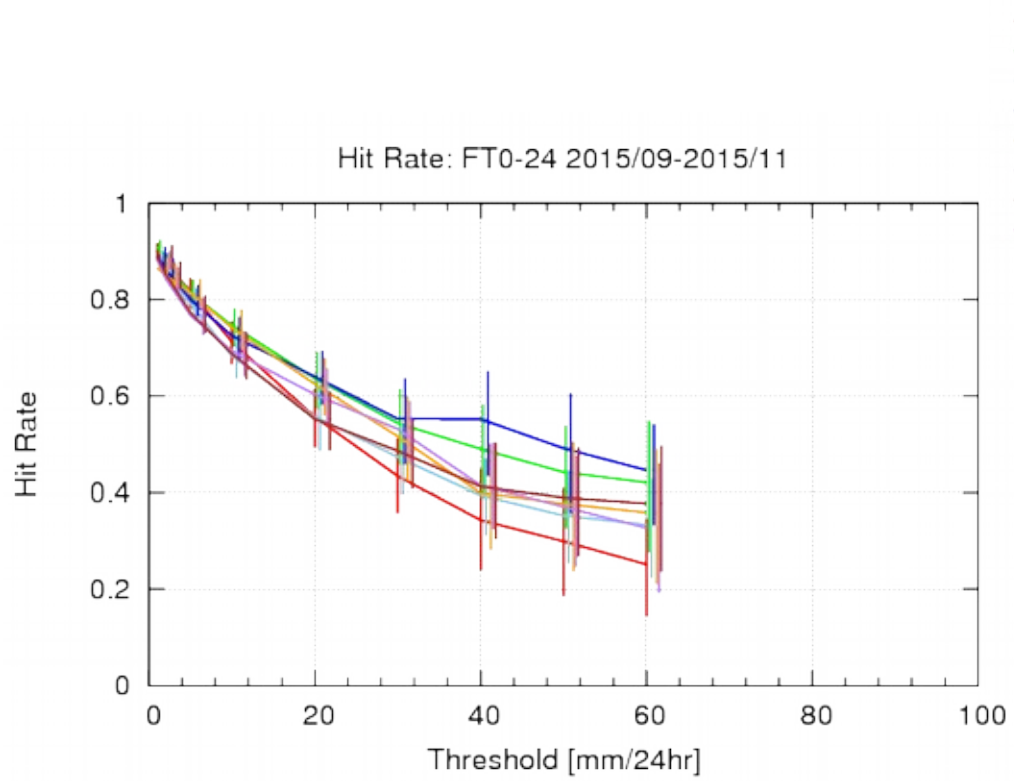


Bias Score: FT0-24 2015/09-2015/11



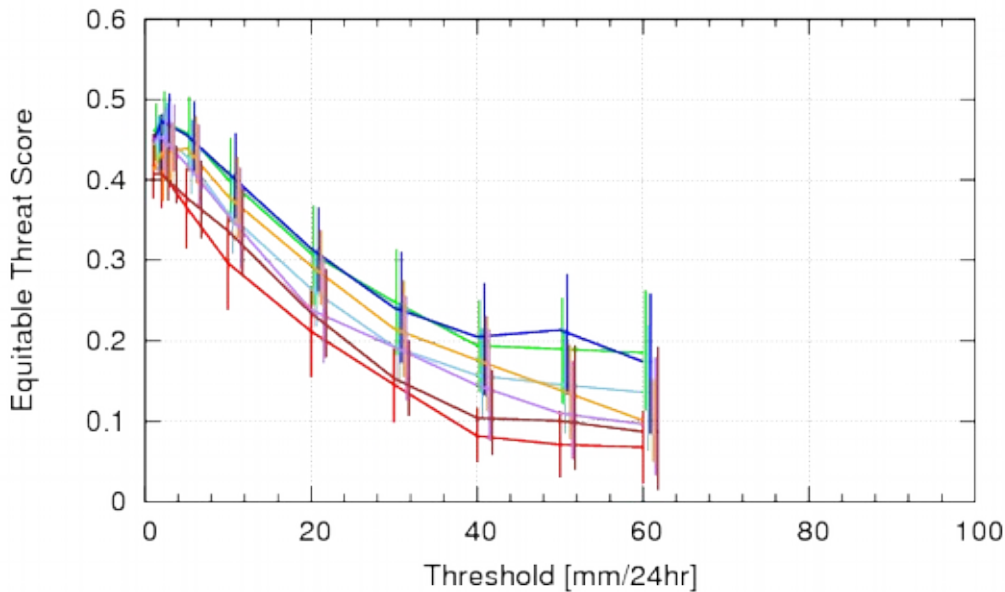
External Dependency Index: FT0-24 2015/09-2015/11



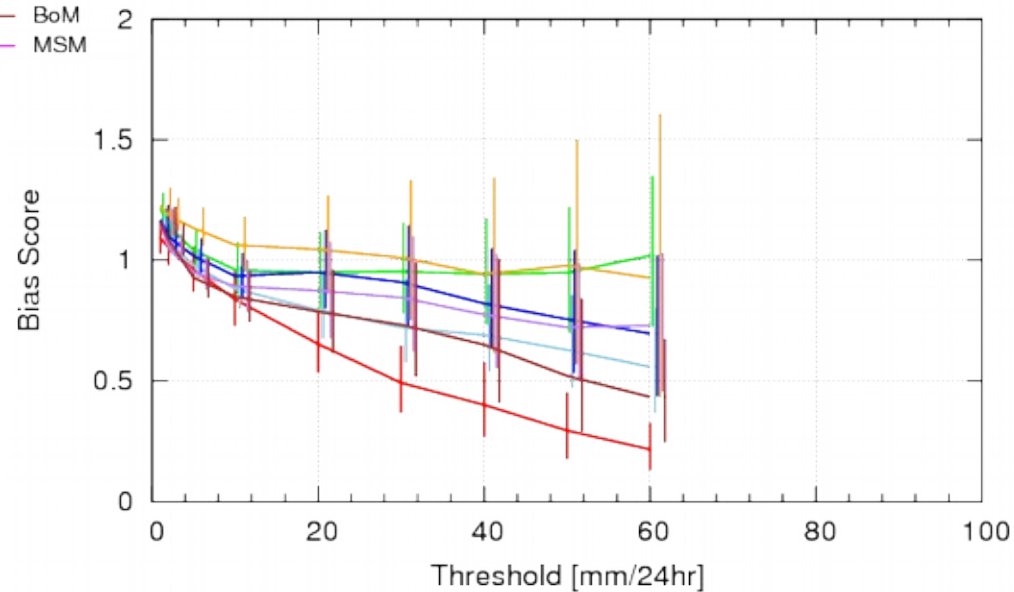


- JMA
- ECMWF
- NCEP
- UKMO
- CMC
- DWD
- BoM
- MSM

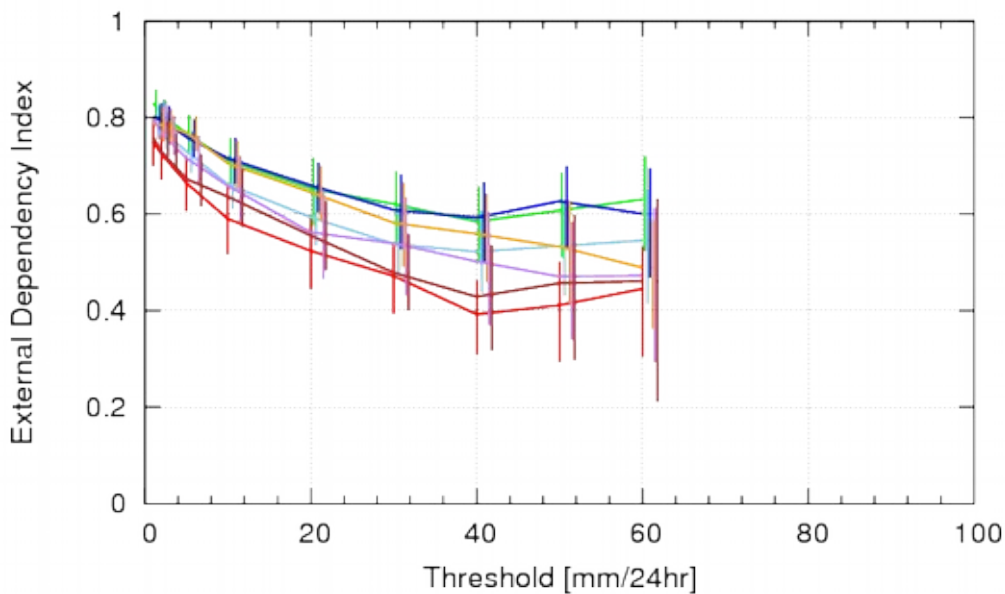
Equitable Threat Score: FT48-72 2015/09-2015/11



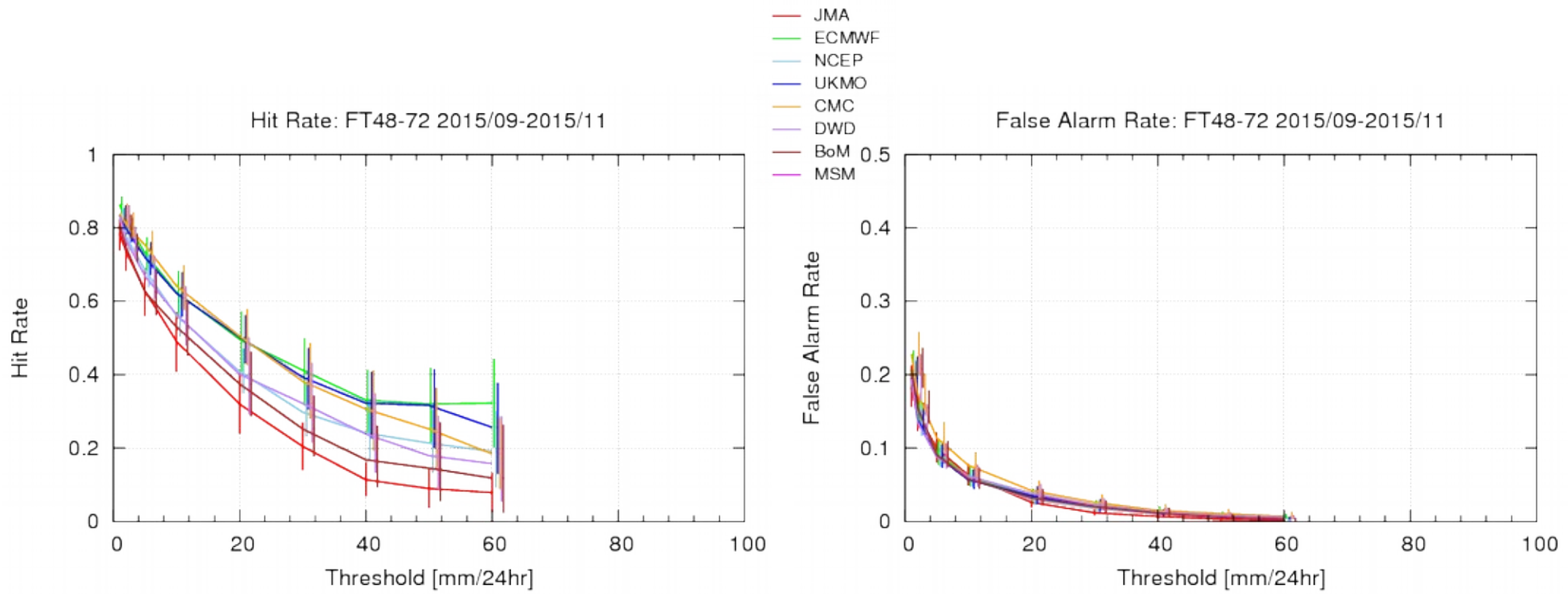
Bias Score: FT48-72 2015/09-2015/11



External Dependency Index: FT48-72 2015/09-2015/11







# Memorandum

# Major upgrades in 2014/12–2015/11

## 2014/11/18

CMC: Introduction of 4D-EnVar (previously 4D-Var). Improved data assimilation. Introduction of new sea-ice concentration analysis.

## 2015/01/14

NCEP: Improved horizontal resolution of forecast from Eulerian T574 (~27 km) to Semi-Lagrangian T1534 (~13 km). Improved horizontal resolution of analysis from Eulerian T192 (~84 km) to Semi-Lagrangian TL574 (~35 km). Improved physics.

## 2015/01/20

DWD: Introduction of ICON with horizontal resolution of 13 km (previously GME with 20 km).

## 2015/05/12

ECMWF: Introduction of Cycle 41r1 (including physics, data assimilation, etc.).

# Major upgrades before 2014/12

## BoM

2012/03/28: improved vertical and horizontal resolution to 40kmL70 to 80kmL50.

## CMC

2013/02/13: Improved horizontal resolution to 25 km from 33 km. Changed vertical coordinate to sigma-p hybrid coordinate from sigma coordinate. Improved 4D-Var data assimilation and model physics.

2013/12/04: Changed the treatment of SST.

## DWD

## ECMWF

2013/06/25: improved vertical resolution to L137 from L91.

2013/11/19: Introduction of Cycle 40r1.

## **NCEP**

## **UKMO**

2014/07/15: Improved horizontal resolution to 17 km from 25 km. Introduction of new dynamical core (ENDGame).

## **JMA**

2007/11/xx: improved horizontal and vertical resolution (20 km, 60 levels, model top 0.1 hPa).

2008/01/xx: Improved deep convection scheme.

2008/08/xx: Improved dynamics. Introduction of adaptive Gauss mesh.

2012/12/xx: Improve Sc scheme.

2013/04/xx: Improved radiation processes.

2014/03/18: Improved the vertical resolution to TL959L100 (top 0.01 hPa) from TL959L60 (0.1 hPa). Improved physics.

## **MSM**



# Major upgrades after 2015/11

## CMC

2015/12/15: GDPS version 5.0.0

## JMA

2016/03/24?: Improved physics.

# Intercomparison over France of QPF from WGNE members models

**Observations : Rain gauges**

**RR24: 24 hours accumulated rainfall**

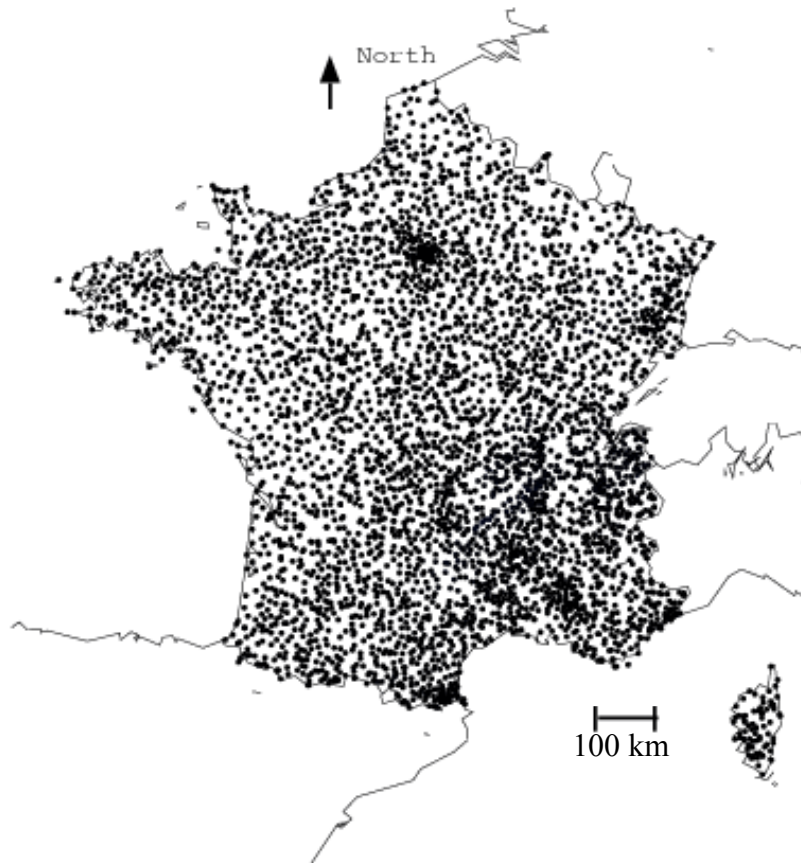
J+1

Bias, FAR, POD and HSS

Thresholds 1mm and 10 mm

# QPF verification

- Average the data and the models QPF at  $0.5^\circ \times 0.5^\circ$



**Climatological state network**

**~4000 raingauges giving 24 hours  
accumulated rain every day**

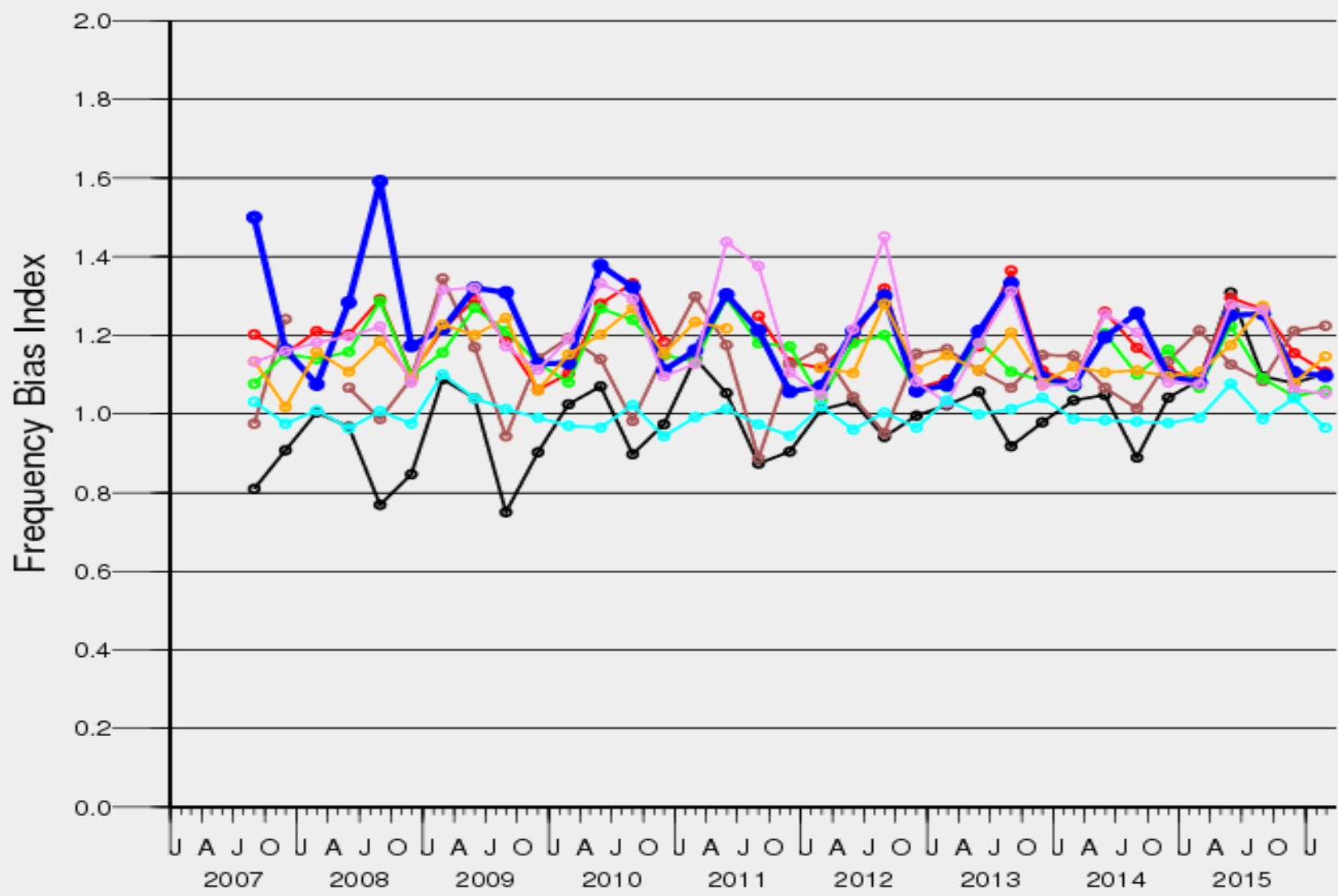



# Frequency Bias Index Precipitation threshold 1 mm/day

Basis 0 UTC, accumulated rainfall 30–54 h, sample common

— DWD — UKMO — CMC — ECMWF — MF — JMA — NCEP — PERS

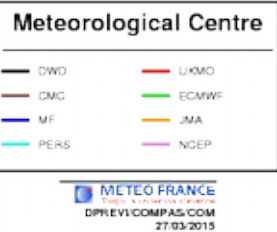
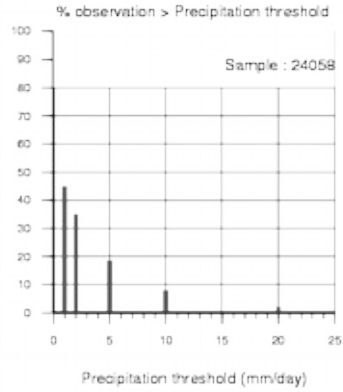
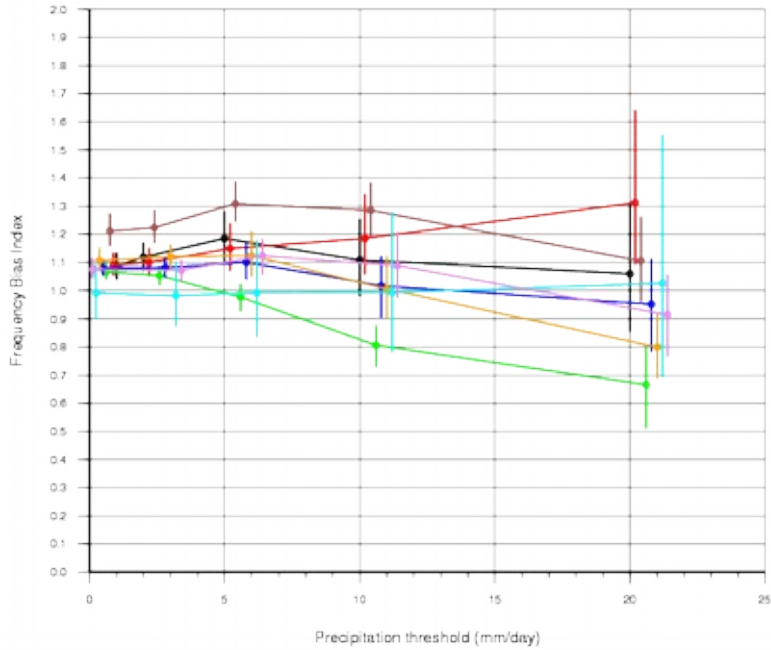
DirOP/COMPAS/COM 30/03/2016





**Frequency Bias Index**  
over 20141201–20150228

basis 0 UTC, accumulated rainfall 30–54 h, sample : common

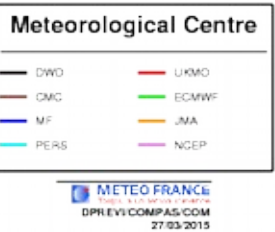
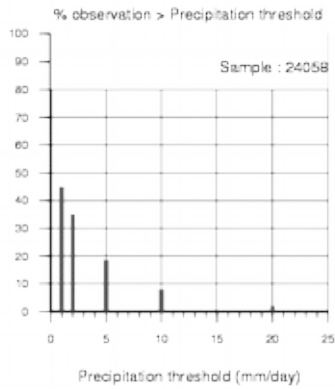
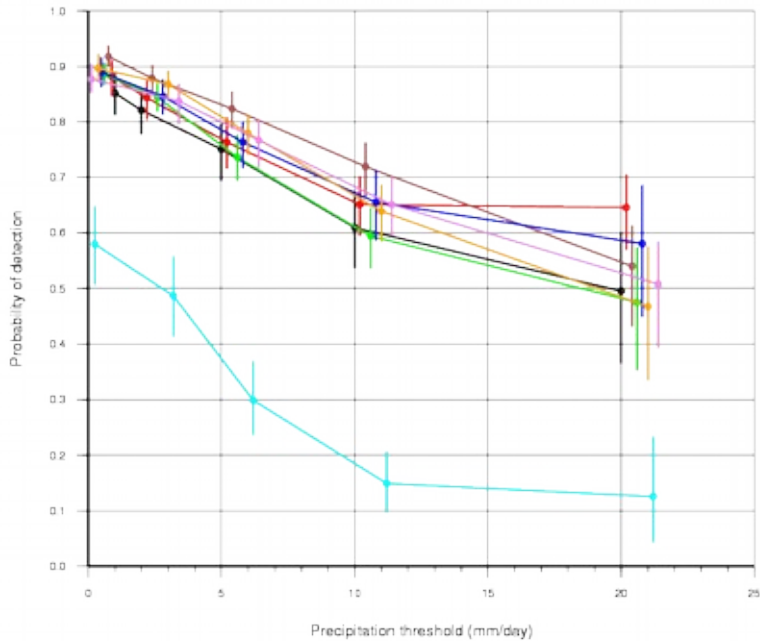


**WINTER**  
**2014-2015**

**Lead time**  
**54 UTC**

**Probability of detection**  
over 20141201–20150228

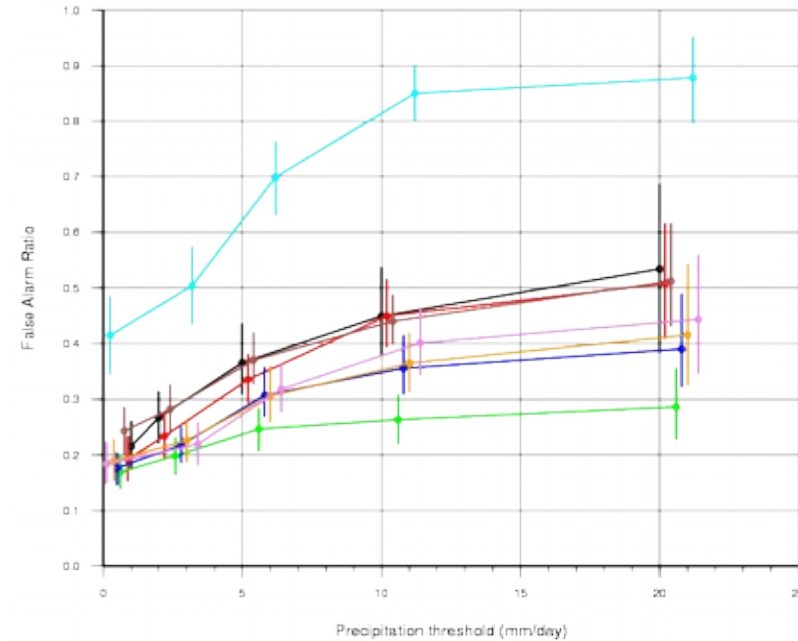
basis 0 UTC, accumulated rainfall 30–54 h, sample : common



**False Alarm Ratio**

over 20141201–20150228

basis 0 UTC, accumulated rainfall 30–54 h, sample : common

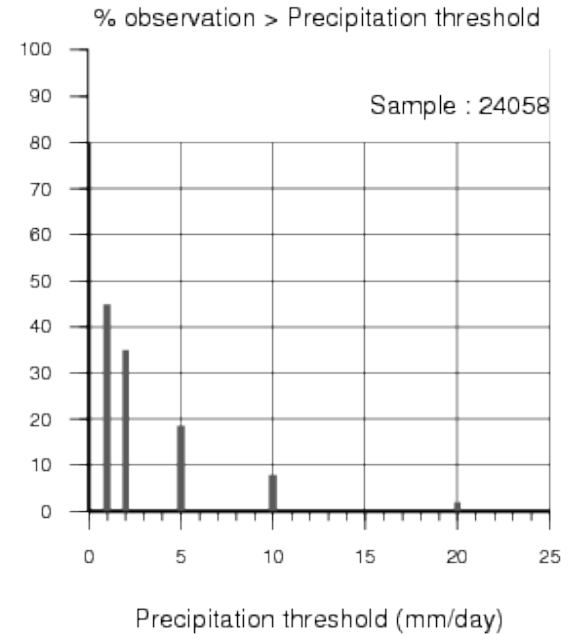
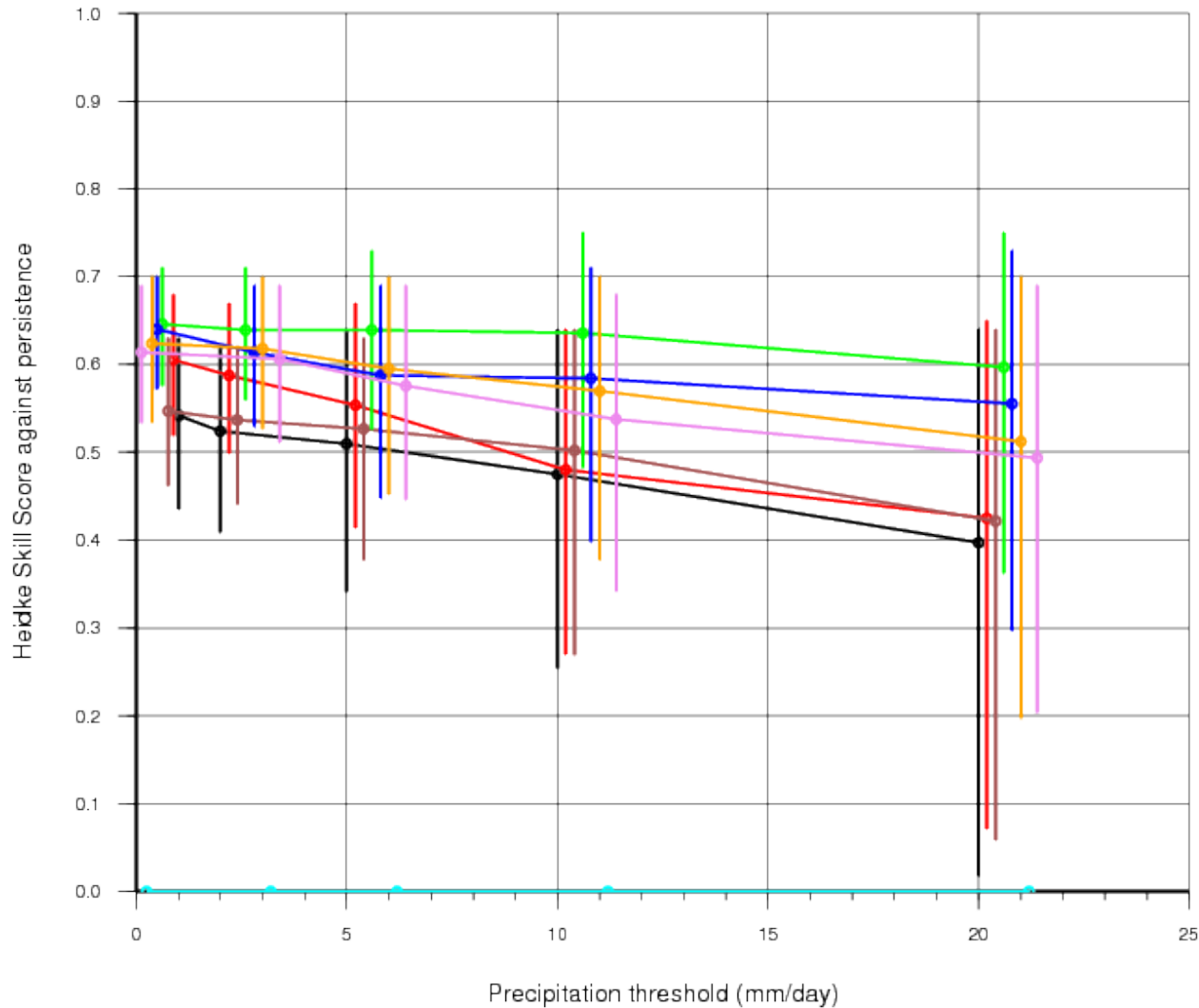


# WINTER 2014-2015

## Lead time 54 UTC

### Heidke Skill Score against persistence over 20141201-20150228

basis 0 UTC, accumulated rainfall 30-54 h, sample : common

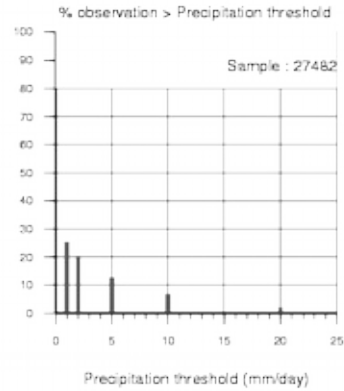
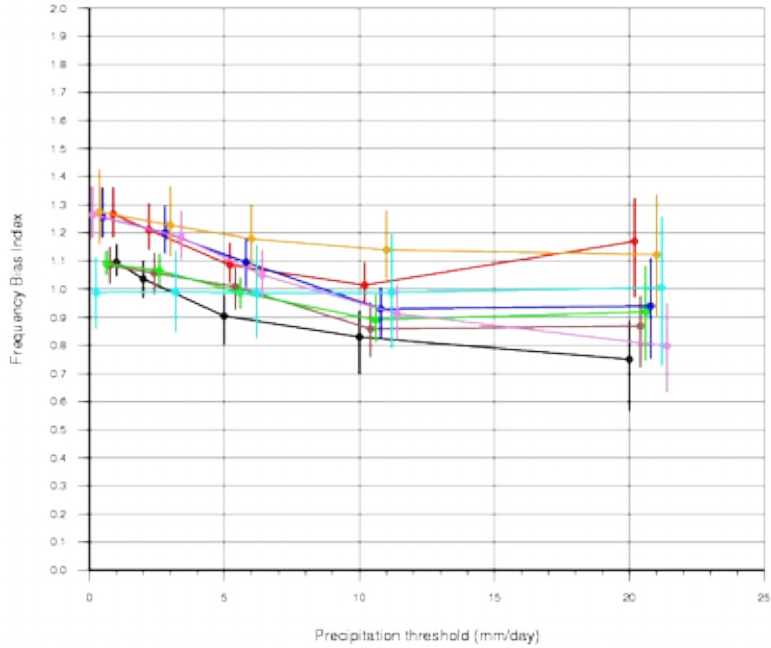


**Meteorological Centre**

— DWD	— UKMO
— CMC	— ECMWF
— MF	— JMA
— PERS	— NCEP

**Frequency Bias Index  
over 20150601–20150831**

basis 0 UTC, accumulated rainfall 30–54 h, sample : common

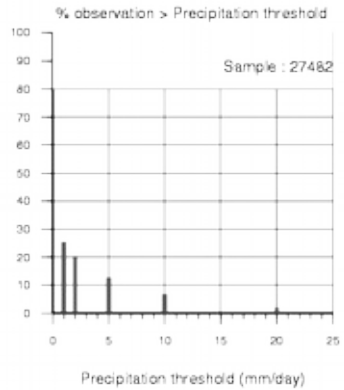
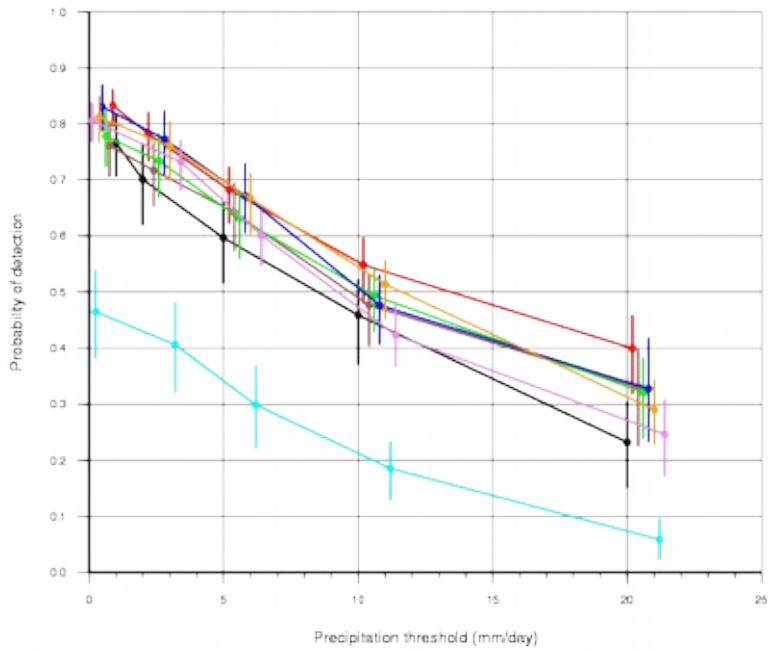


**Summer  
2015**

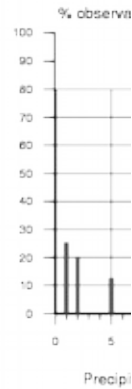
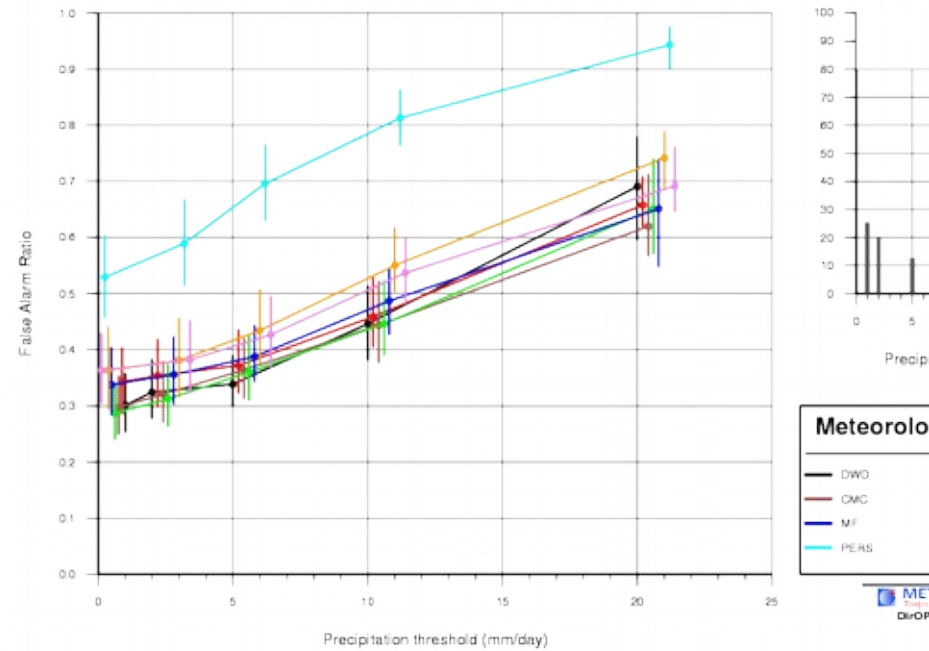
**Lead time  
54 UTC**

**Probability of detection  
over 20150601–20150831**

basis 0 UTC, accumulated rainfall 30–54 h, sample : common

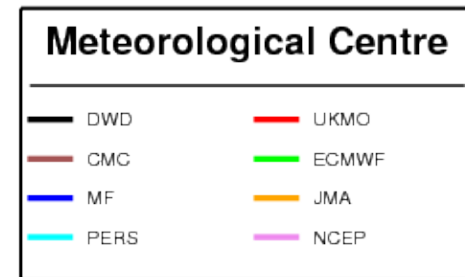
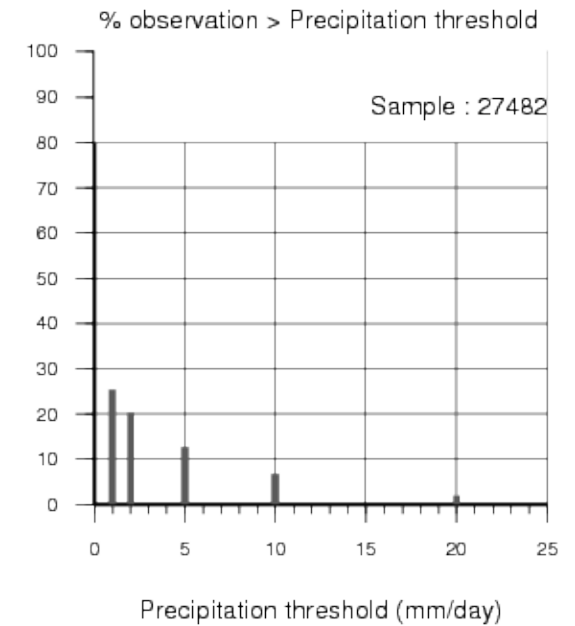
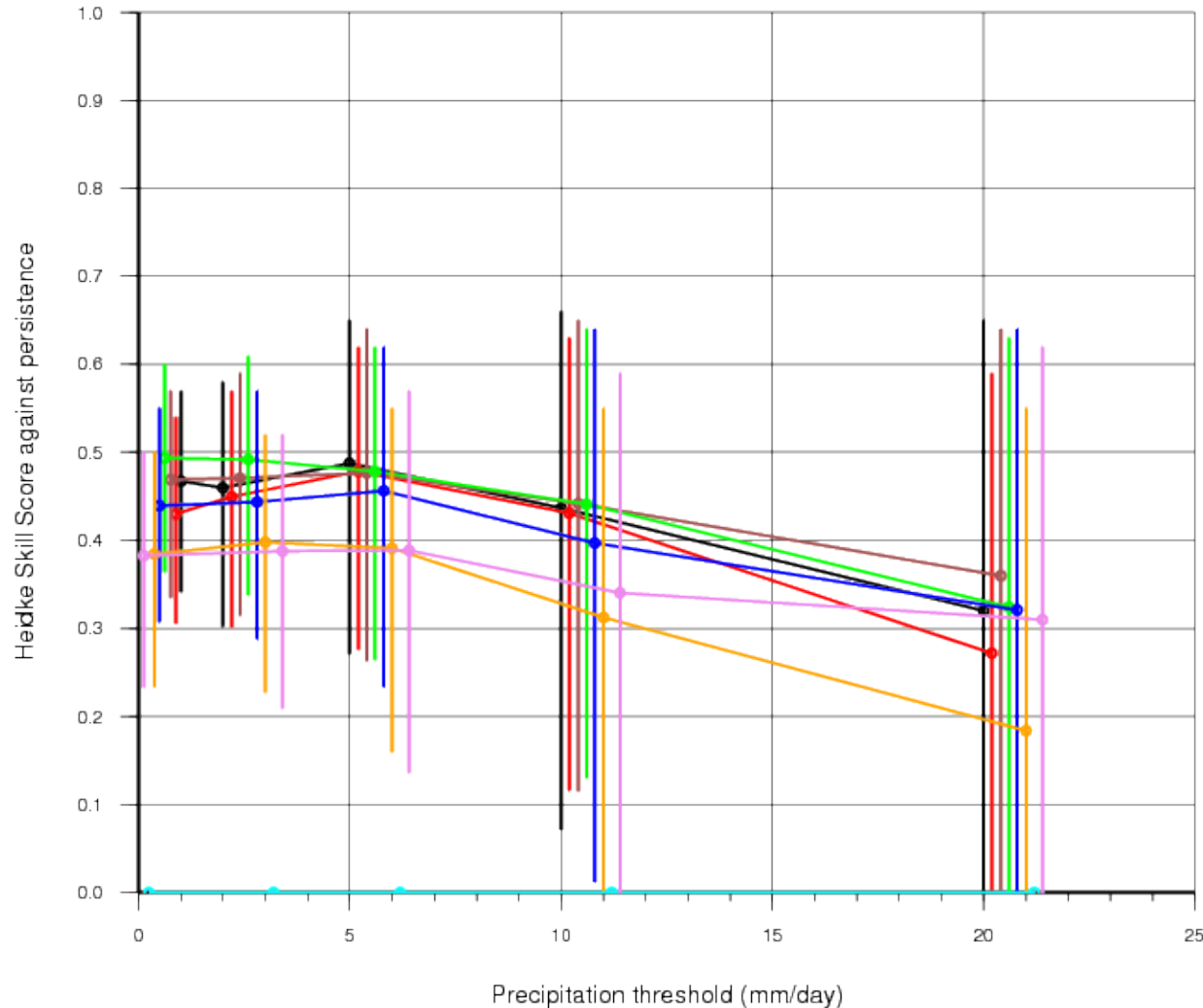


**False Alarm Ratio  
over 20150601–20150831**  
basis 0 UTC, accumulated rainfall 30–54 h, sample : common



## Heidke Skill Score against persistence over 20150601–20150831

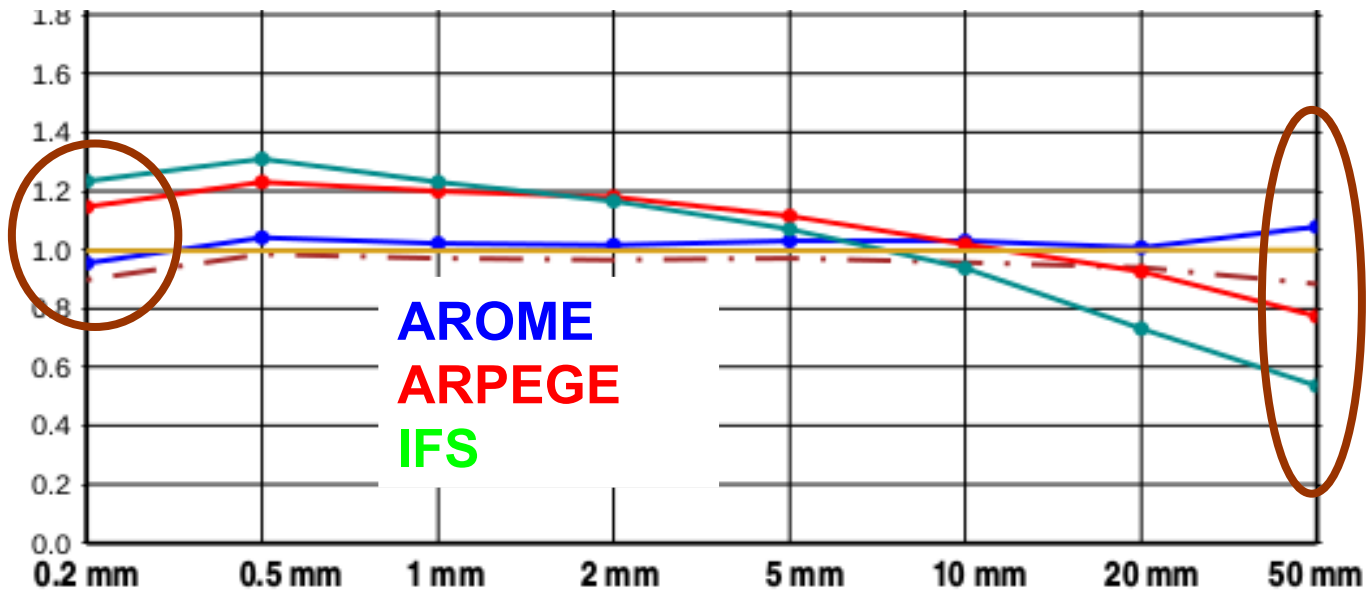
basis 0 UTC, accumulated rainfall 30–54 h, sample : common



# **QPF verification over France from operational models against high resolution observations**

- **Gridded observations : combined radar-gauge analyses (ANTILOPE)**
- **RR24**
- Verification grid  $0.025^\circ$
- Year 2014
- Bias, BSS\_NO
- Thresholds 0.2, 0.5, 1, 2, 10, 20 and 50 mm

# RR24 Year 2015 Frequency Bias

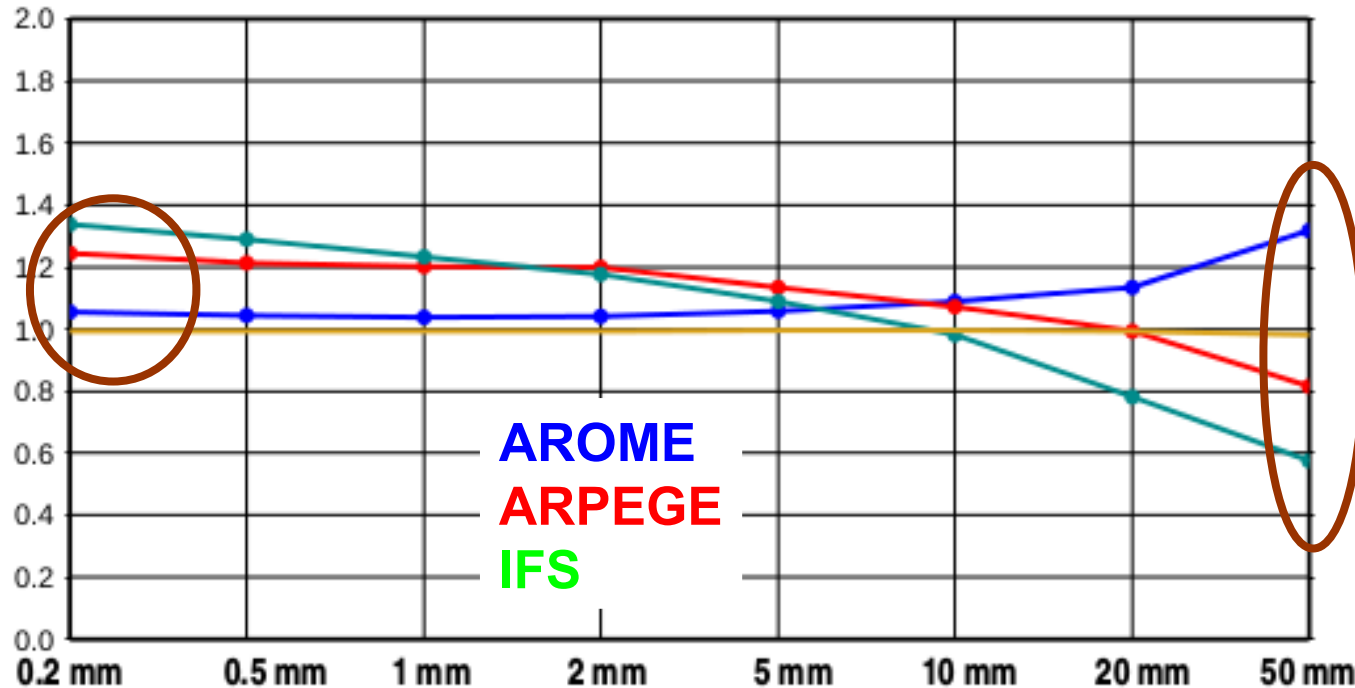


Against Rain gauges  
Stations [ 4000 pts

ANTILOPE BIAS :

Slight  
underestimation for  
weak rains

10% underestimation  
rr24 >= 50mm/24H



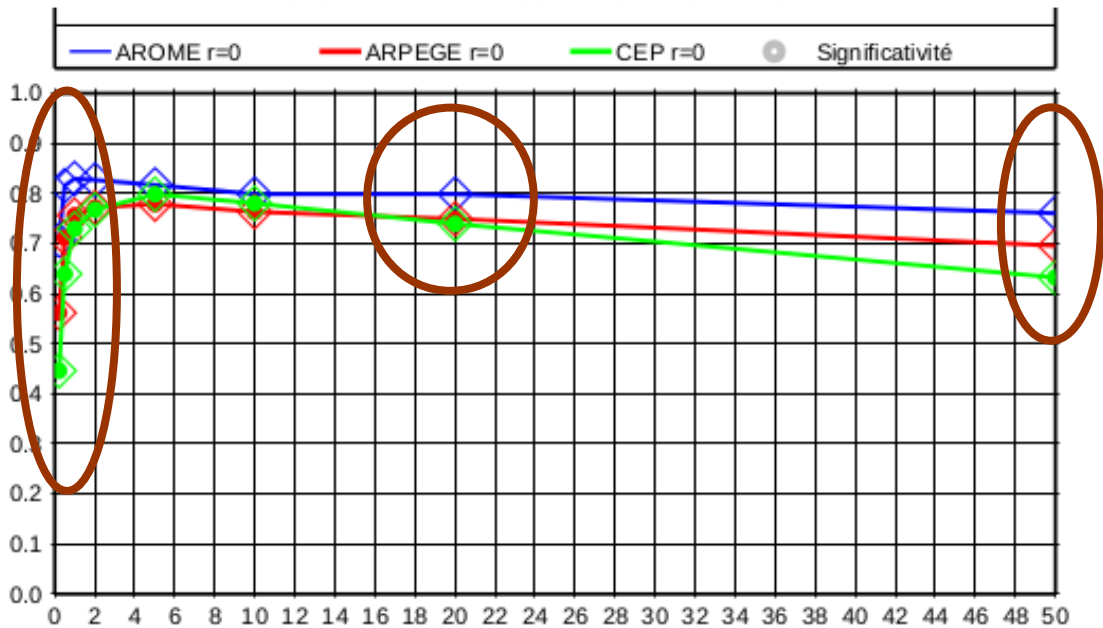
Against ANTILOPE  
[ 12500 pts

**AROME bias for heavy  
rains increases when  
ANTILOPE  
underestimates  
phenomena**



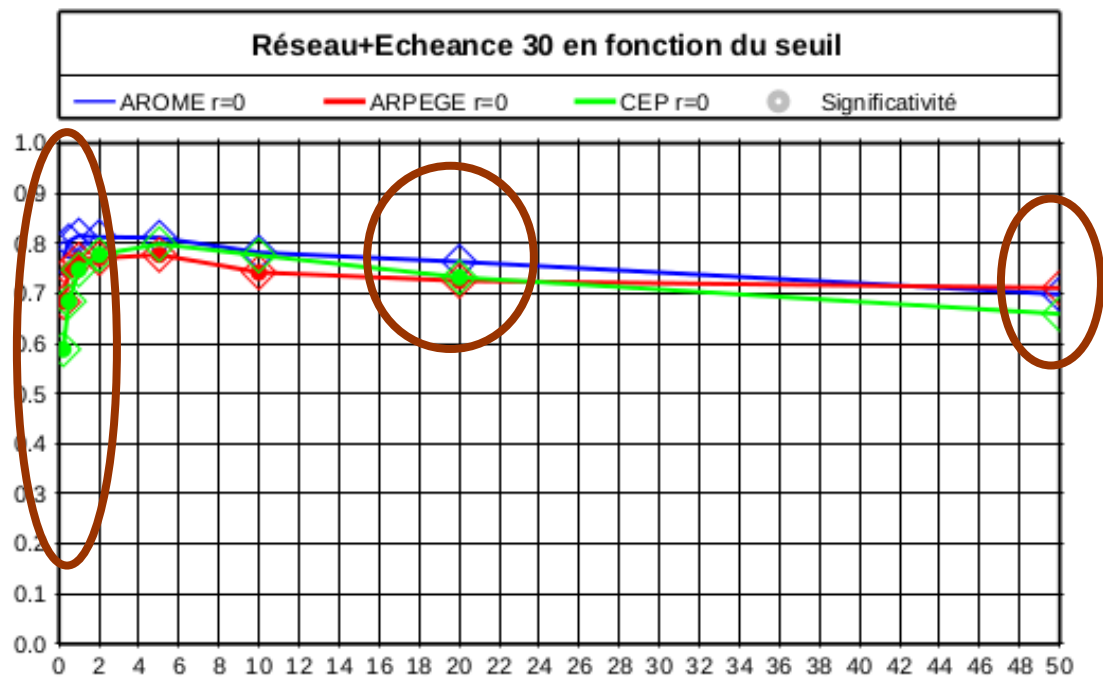
# RR24 Year 2015

# BSS\_NO 50 km



Against Rain gauges stations

AROME  
ARPEGE  
IFS

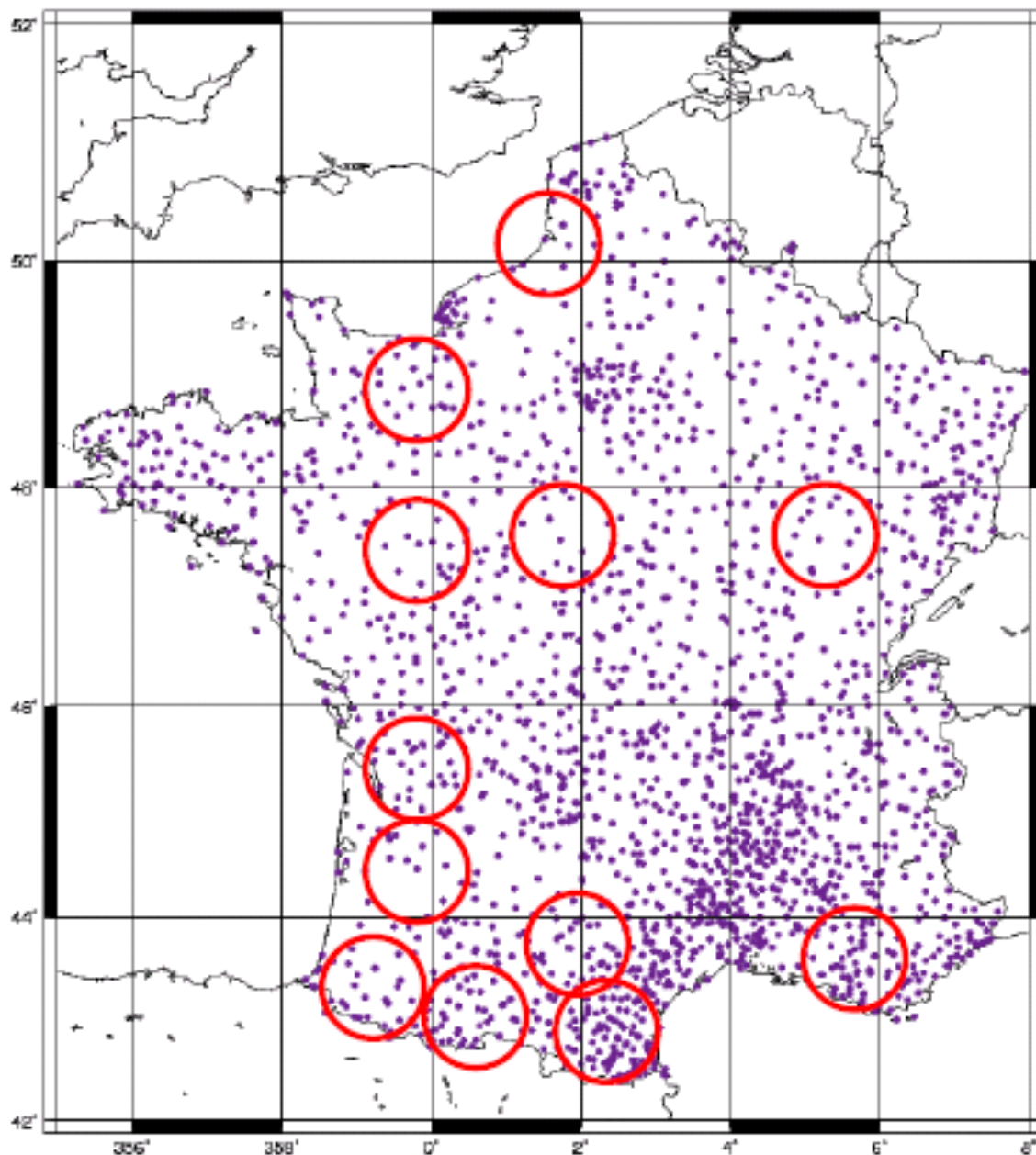


Against ANTILOPE :  
Better estimation of rain/no rain threshold  
similar results with both references except for 50mm threshold

# **QPF verification over France from the operational HR model AROME using neighbourhood method**

- **RR6: 6 hours accumulated rainfall**
- **recommandation : FSS**
- **Météo-France choice : BSS**

# 6 hours accumulated rainfall



data and models QPF are averaged on  $0.0025^\circ$  squares

Climatological state network

~1800 raingauges giving hourly accumulated rain

Red circles of radius 50 km give exemples of neighbourhood

# FSS versus BSS\_NO

$$BS \text{ or } FBS = \frac{1}{N_{days}} \sum_{j=1}^{N_{days}} \frac{1}{N_{obs}} \sum_{o=1}^{N_{obs}} \left( v_{forecast}(o, j) - v_{obs}(o, j) \right)^2$$

$$FSS = 1 - \frac{FBS}{FBS \text{ reference}}$$

$$BSS\_NO = 1 - \frac{FBS}{FBS \text{ persistence}}$$

$$FBS \text{ reference} = \frac{1}{N_{days}} \sum_{j=1}^{N_{days}} \frac{1}{N_{obs}} \left[ \sum_{o=1}^{N_{obs}} v_{forecast}(o, j)^2 + \sum_{o=1}^{N_{obs}} v_{obs}(o, j)^2 \right]$$

# 6 hours accumulated rainfall BSS\_NO

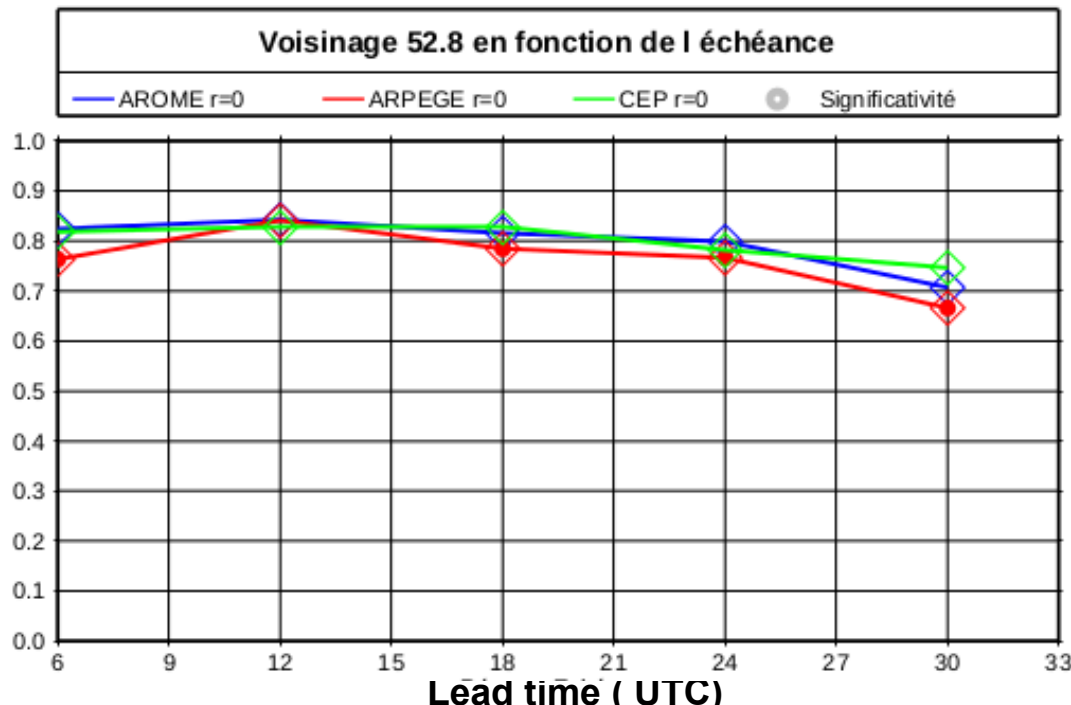
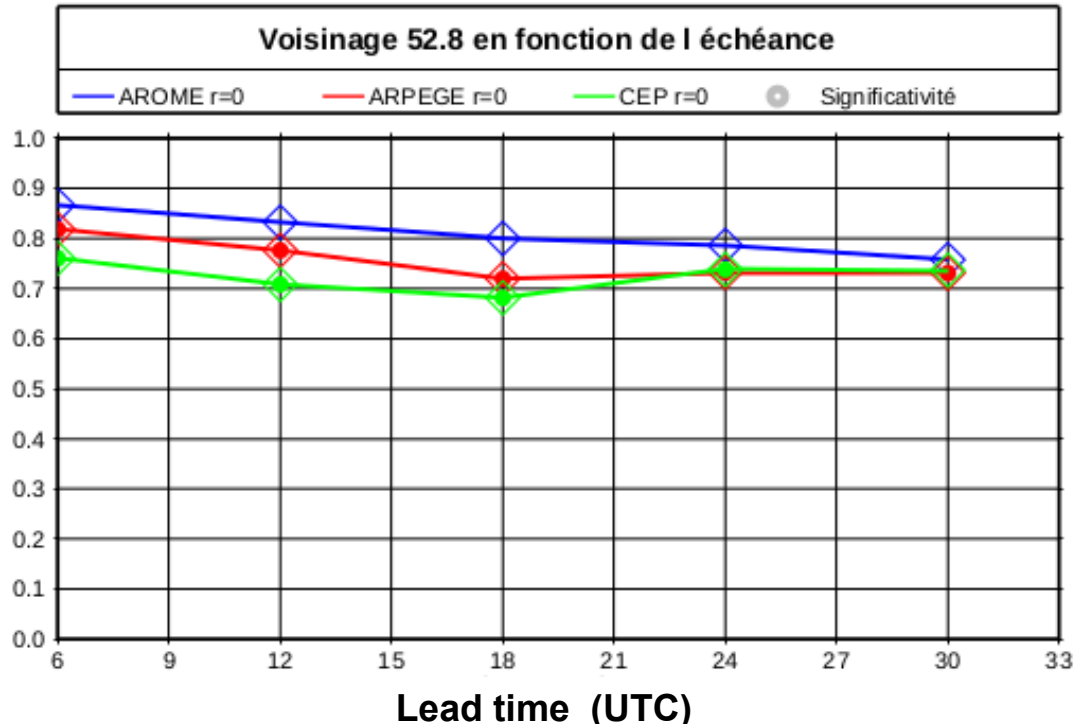
Neighbourhood **50 km**

Winter 2014-2015

Threshold **0.5 mm**

**AROME**  
**ARPEGE**  
**IFS**

Threshold **5 mm**

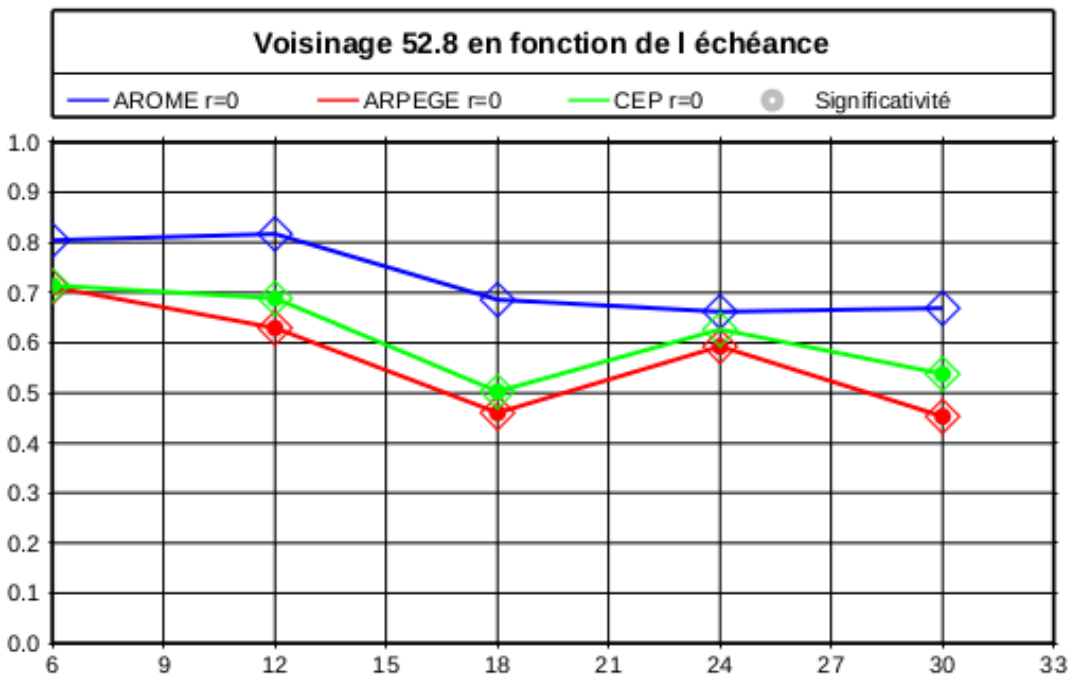
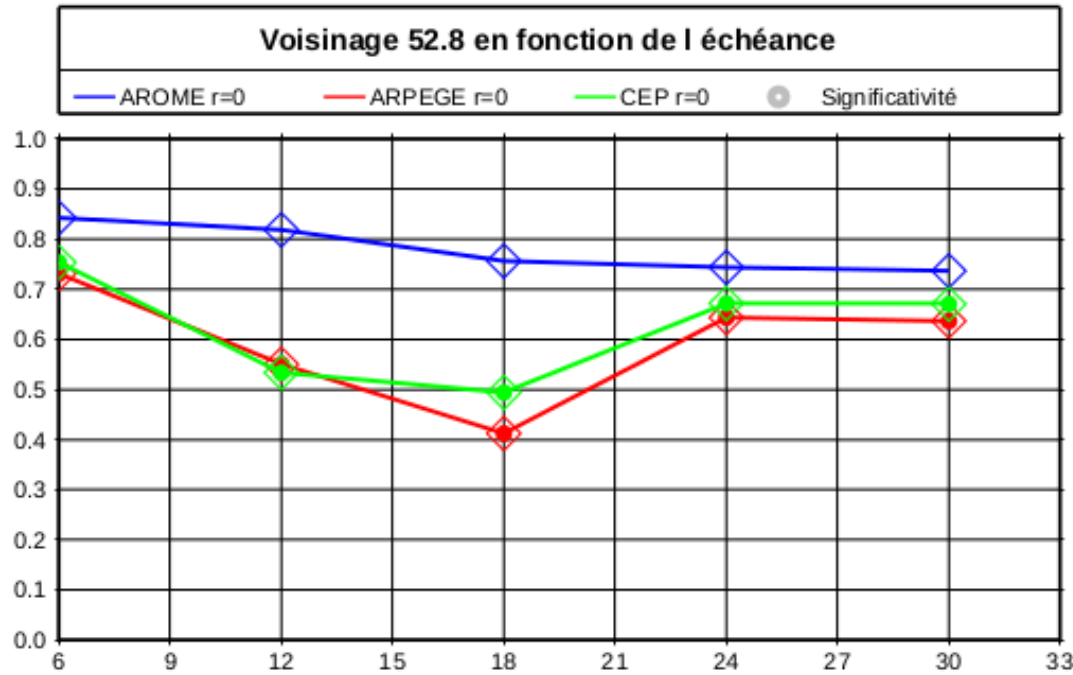


# 6 hours accumulated rainfall BSS\_NO

Neighbourhood 50 km  
Summer 2015

Threshold 0.5 mm

Threshold 5 mm



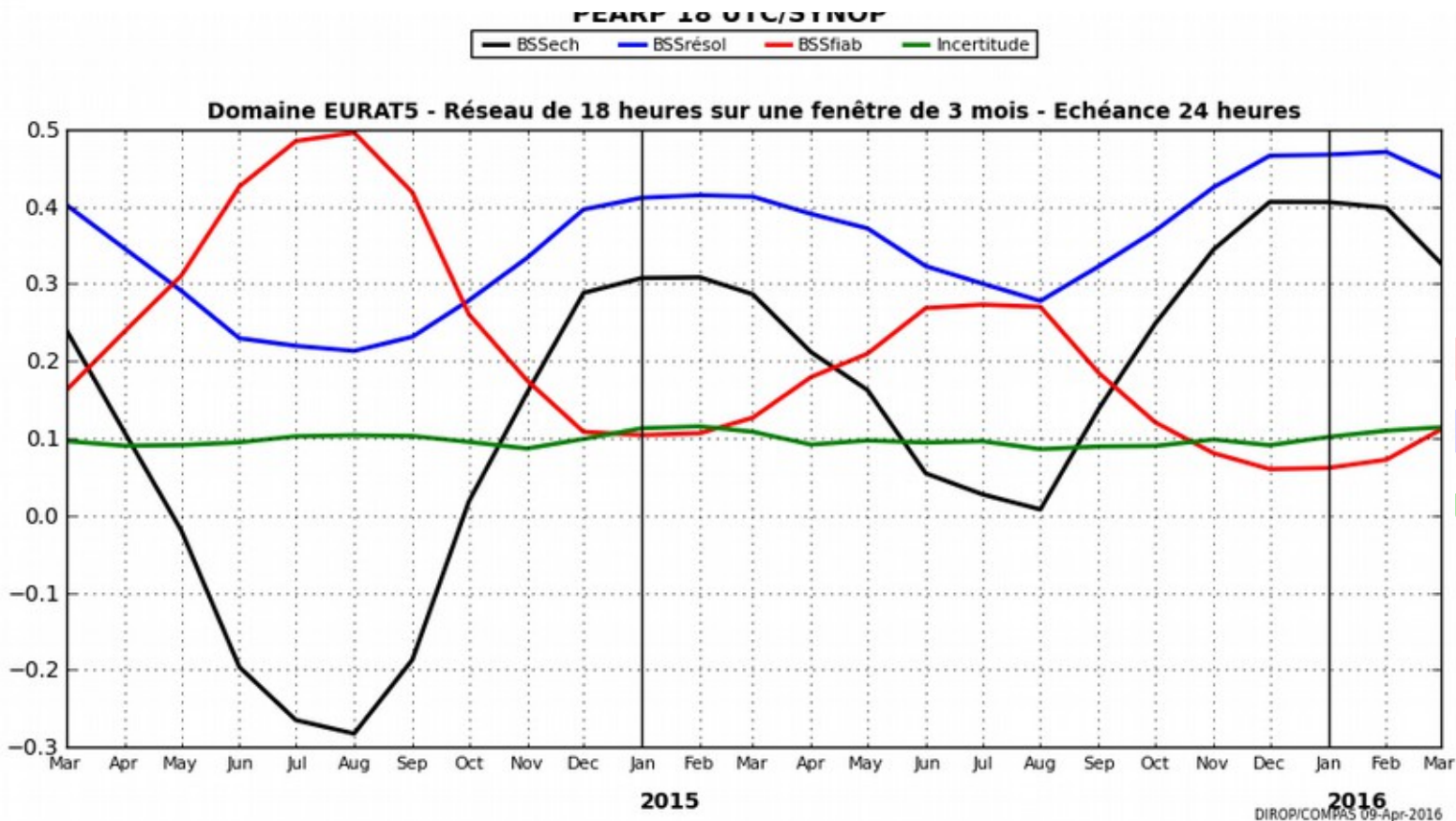


# Probabilistic forecasts from ensembles

- RR6 verification of PEARP
  - ✓ for thresholds 0.2 1 2 et 4mm/6h // raingauges ;
  - ✓ 0.5° grid
  - ✓ Verification using the nearest grid point of the observation
- RR24 verification for thresholds 1 5 10 20mm /24h
- Scores :
  - ✓ BSS and components
  - ✓ Reliability diagram
  - ✓ Roc diagram and roc area

# BSS and components

PEARP 18 UTC/SYNOP      Domain EURAT5  
6 hours accumulated rainfall for threshold 1 mm

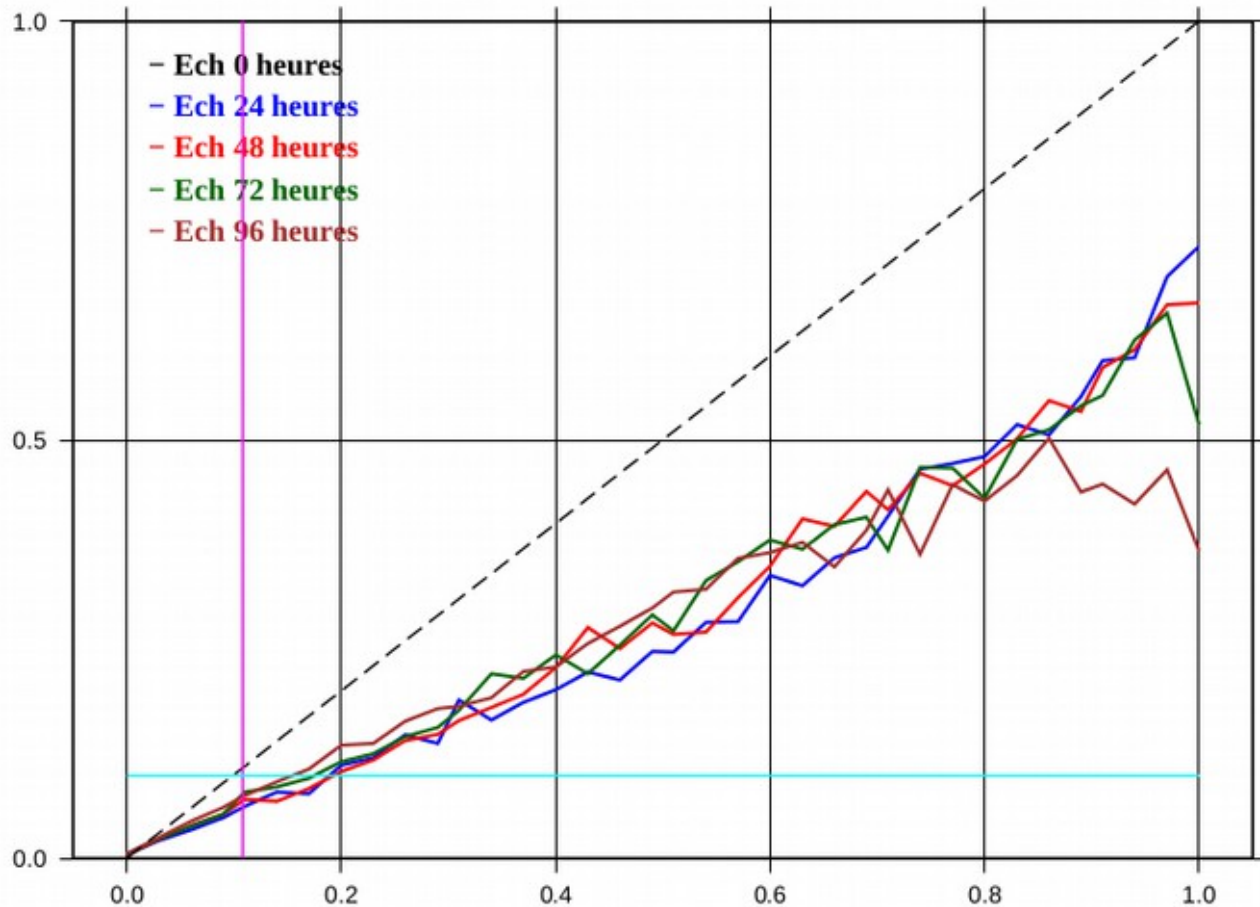


**BSS\_clim**  
**BS\_reliability**  
**BS\_resolution**  
**uncertainty**

# Reliability diagram

PEARP/SYNOP

Domain EURAT5



**SUMMER 2015**

6 hours accumulated rainfall  
threshold 1 mm  
Issu 18UTC  
validity 18 UTC

# Roc diagram and roc area

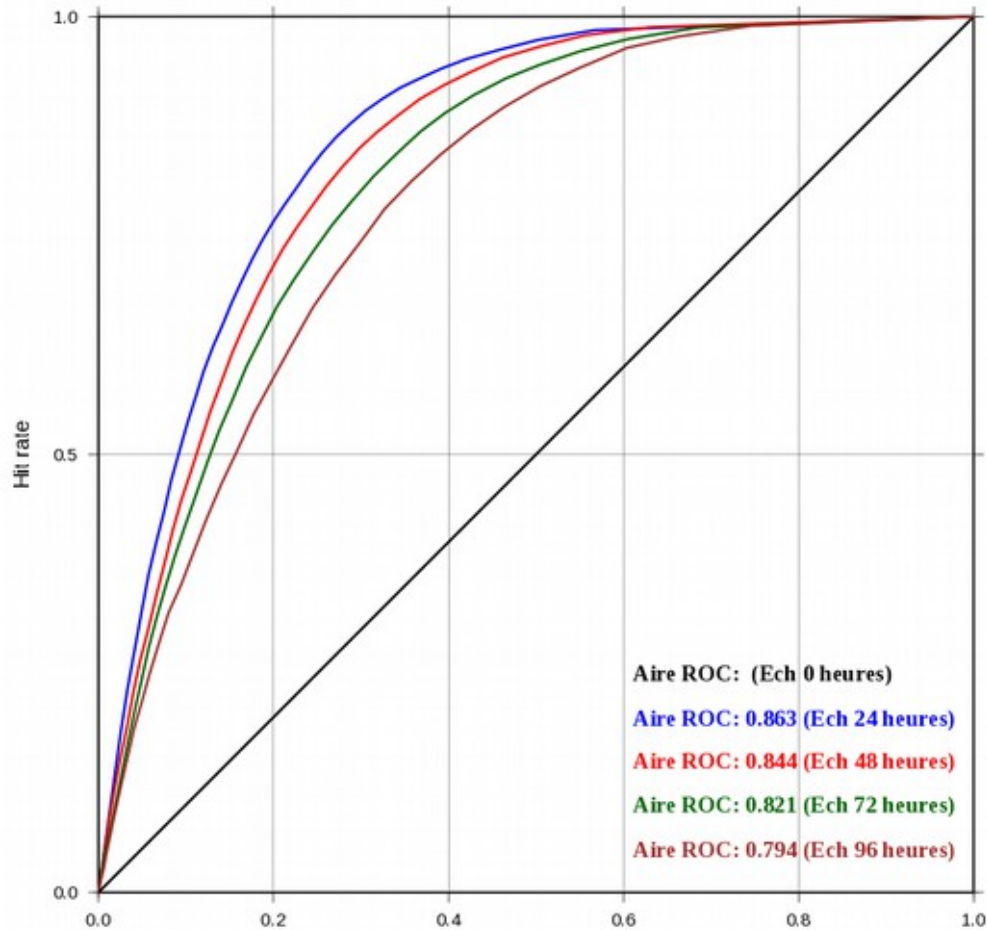
PEARP/SYNOP

Domain EURAT5

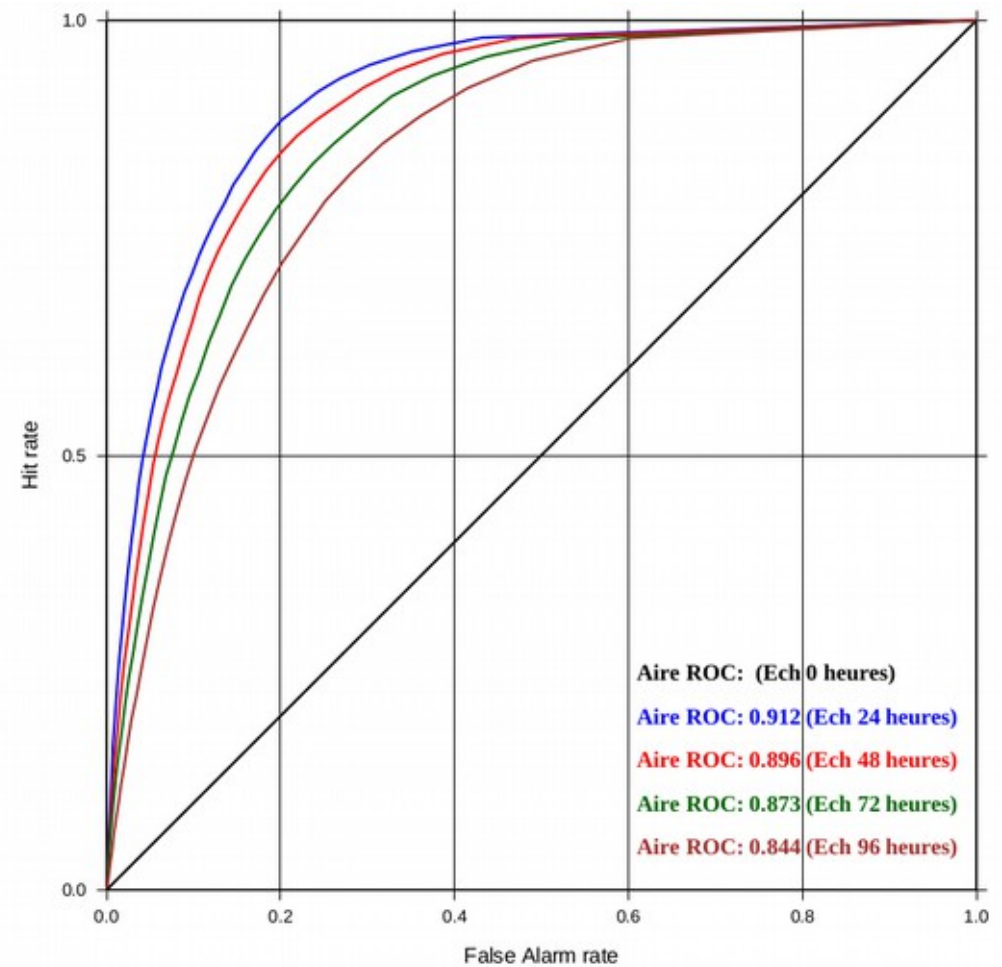
6 hours accumulated rainfall

threshold 1 mm / issu 18 UTC / validity 18 UTC

Summer 2014



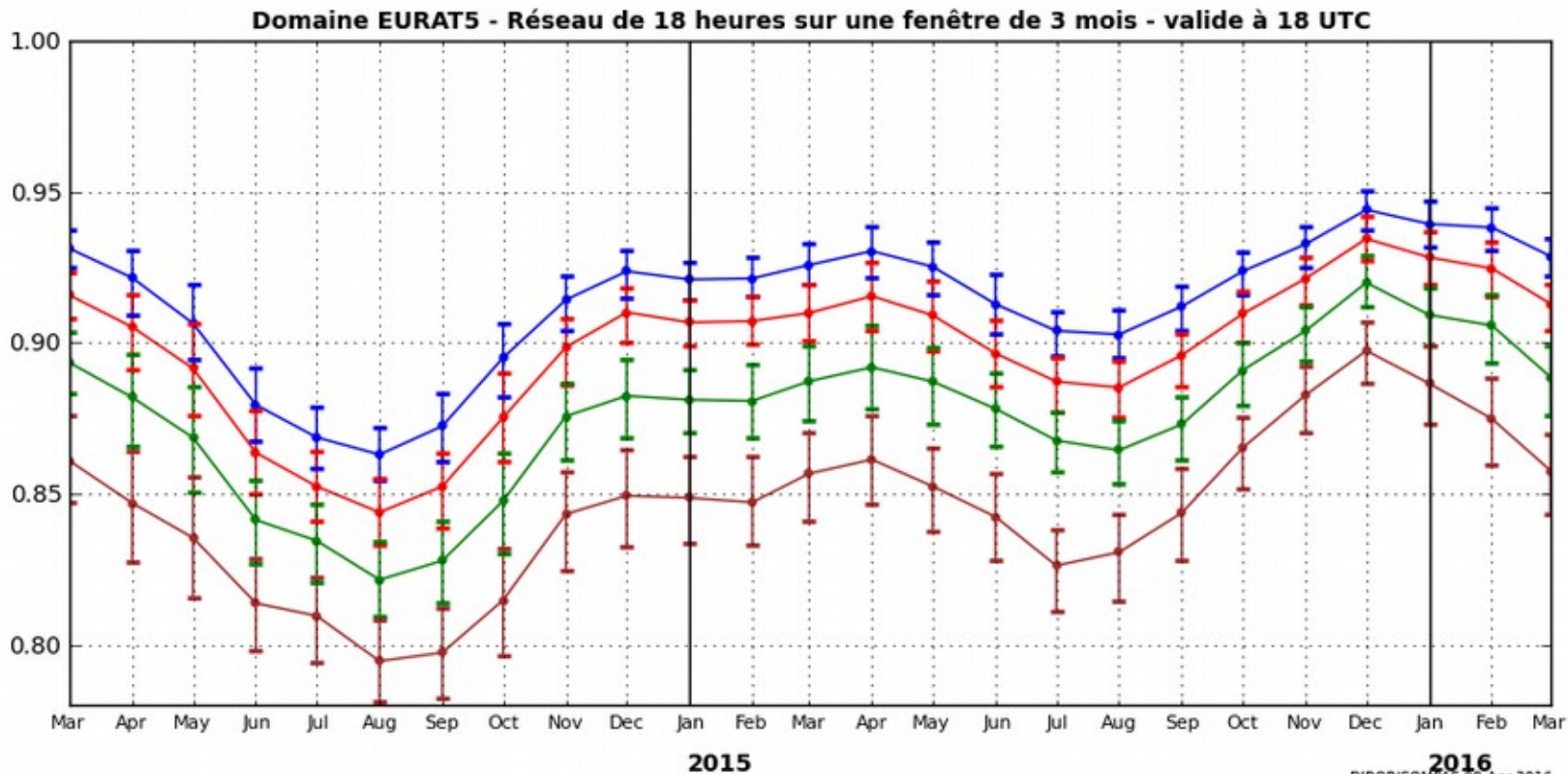
SUMMER 2015



# Roc area : 3 month mean evolution

PEARP 18 UTC/SYNOP      Domain EURAT5  
6 hours accumulated rainfall for threshold 1 mm

des précipitations en 6 heures pour un seuil de 1 mm  
PEARP 18 UTC/SYNOP



24 h range  
48 h range  
72 h range  
96 h range  
& quantiles 5 95%

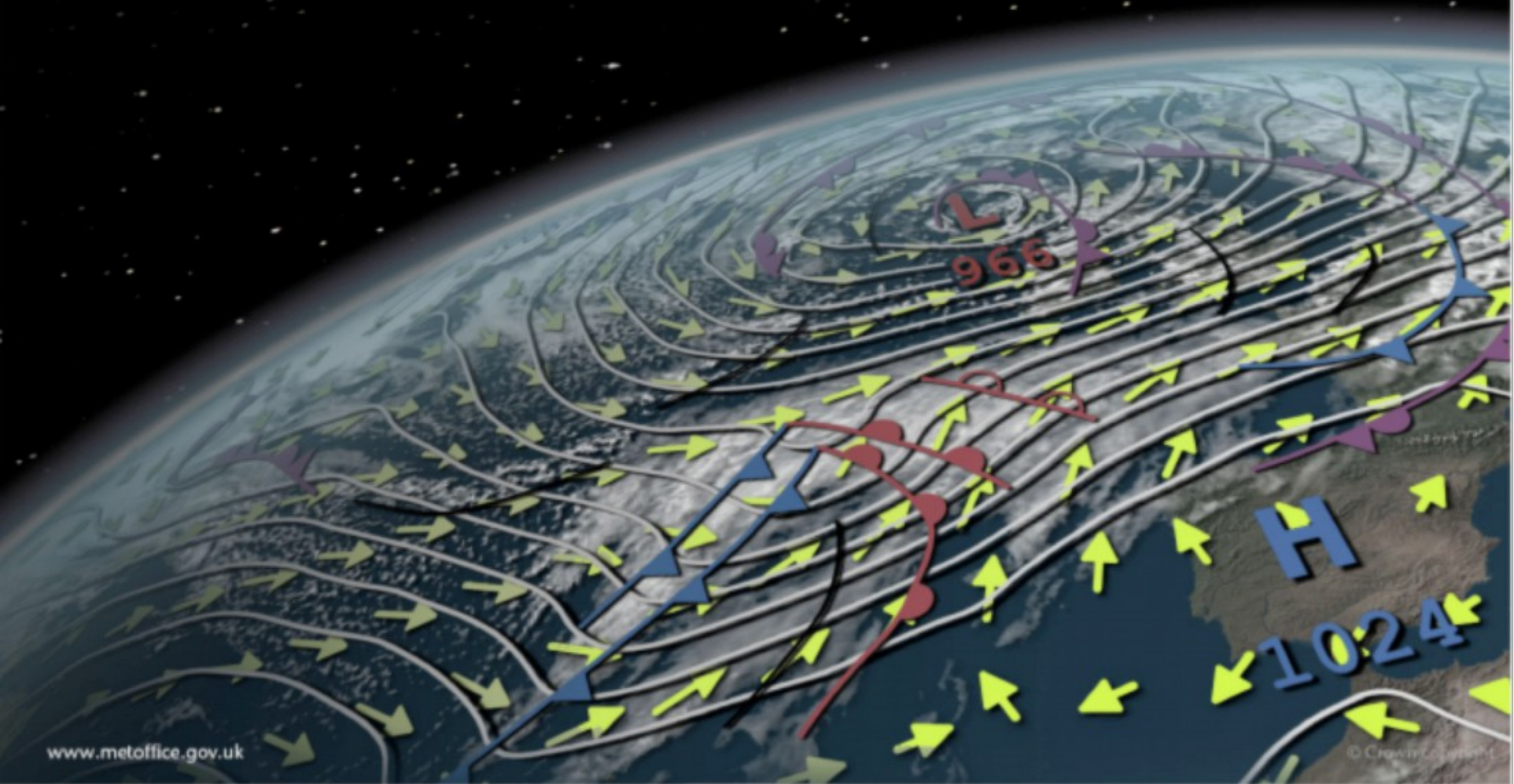
**UKMO**





Met Office

# Global model precipitation verification



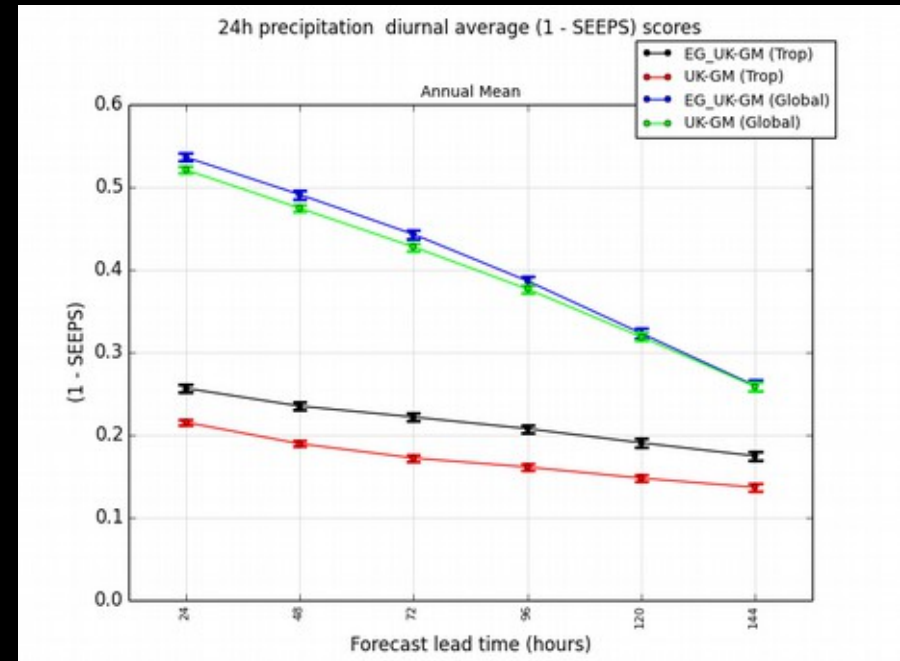
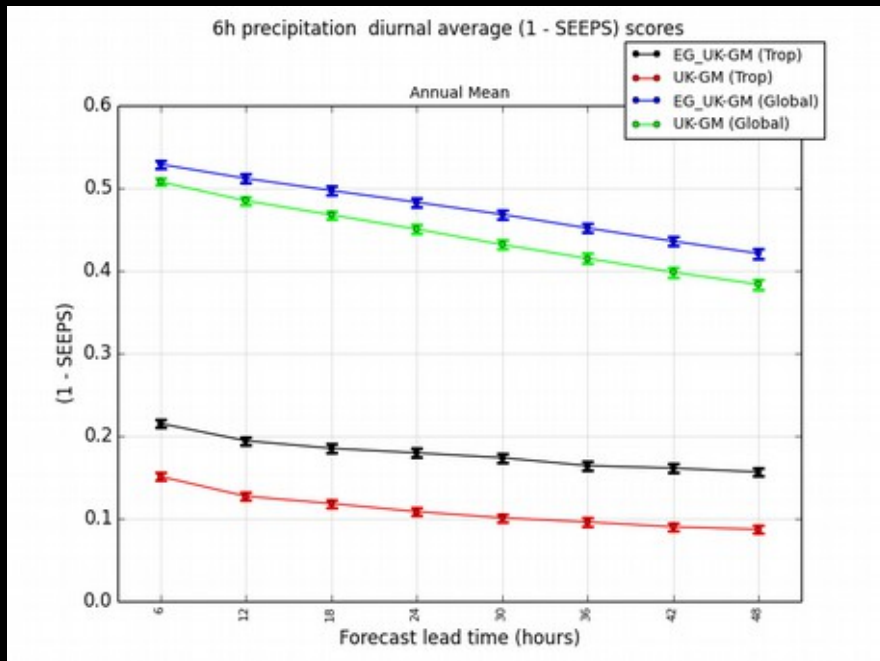
# Verification of QPF using SEEPS

Score with forecast lead time, April 2012 to February 2016

SEEPS skill score from UM  
Global  
6-hour accumulations  
(6h to 48h)



Diurnal averages  
Tropics: red & black  
Global: blue & green



SEEPS skill score from UM  
Global  
24-hour precipitation  
accumulations (day 1 to 6)

# Verification of QPF using SEEPS

Score with forecast lead time, April 2012 to February 2015

- Globally, model has useful skill.
- SEEPS shows model has almost 3x skill globally than in the Tropics.
- Latest UM upgrades (labelled “EG\_” in previous slide) confirm signs of improvement over the Tropics.
- Tropics errors are almost constant with forecast lead time.



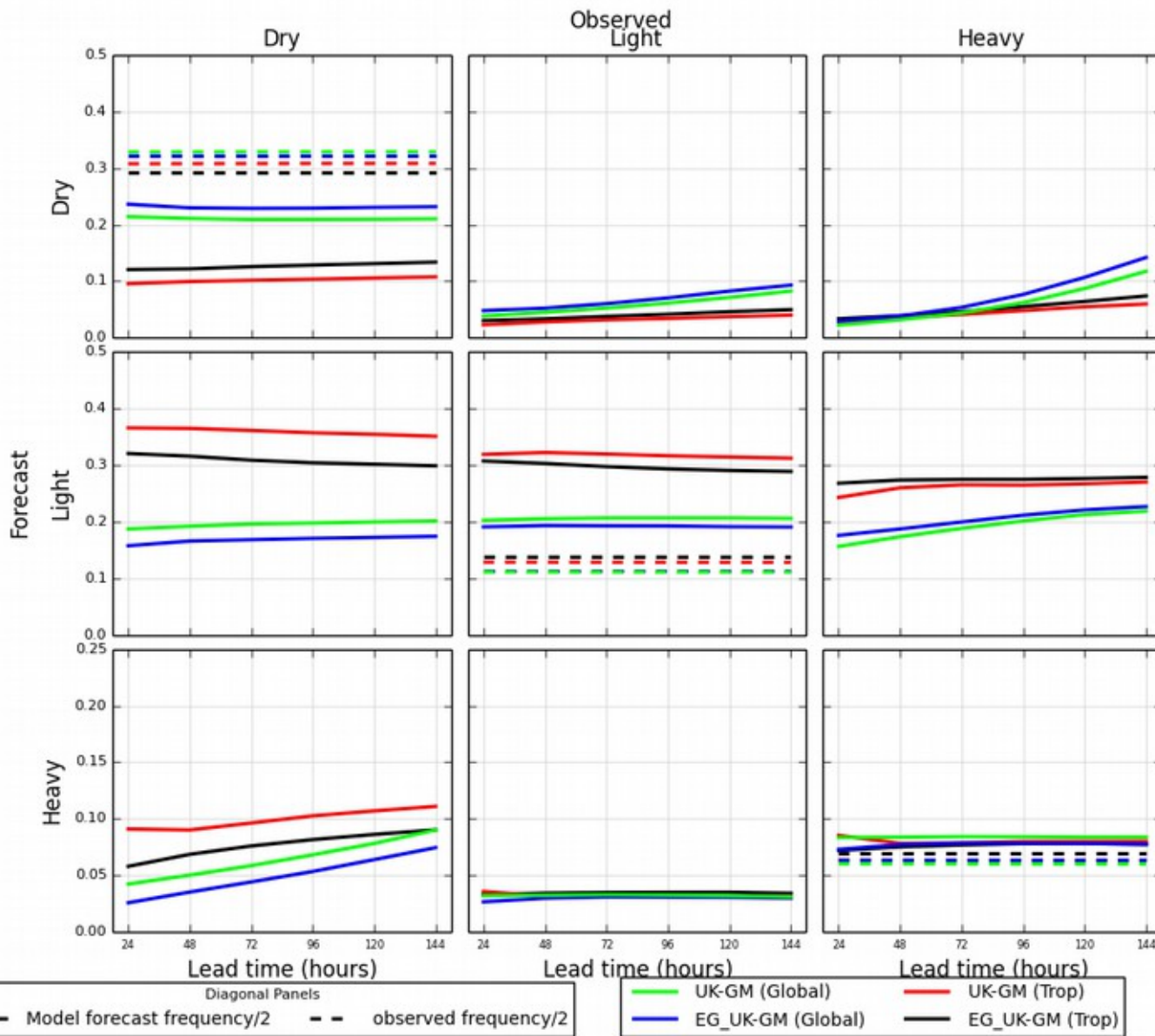
# Verification of QPF using SEEPS

## Decomposition into constituent error sources

Diurnal  
Average  
2012-2016

24-hour  
totals

UK-GM, EG\_UK-GM diurnal average SEEPS decomposition  $S_{ef}$  trial average over dates (20120401 to 20160229)



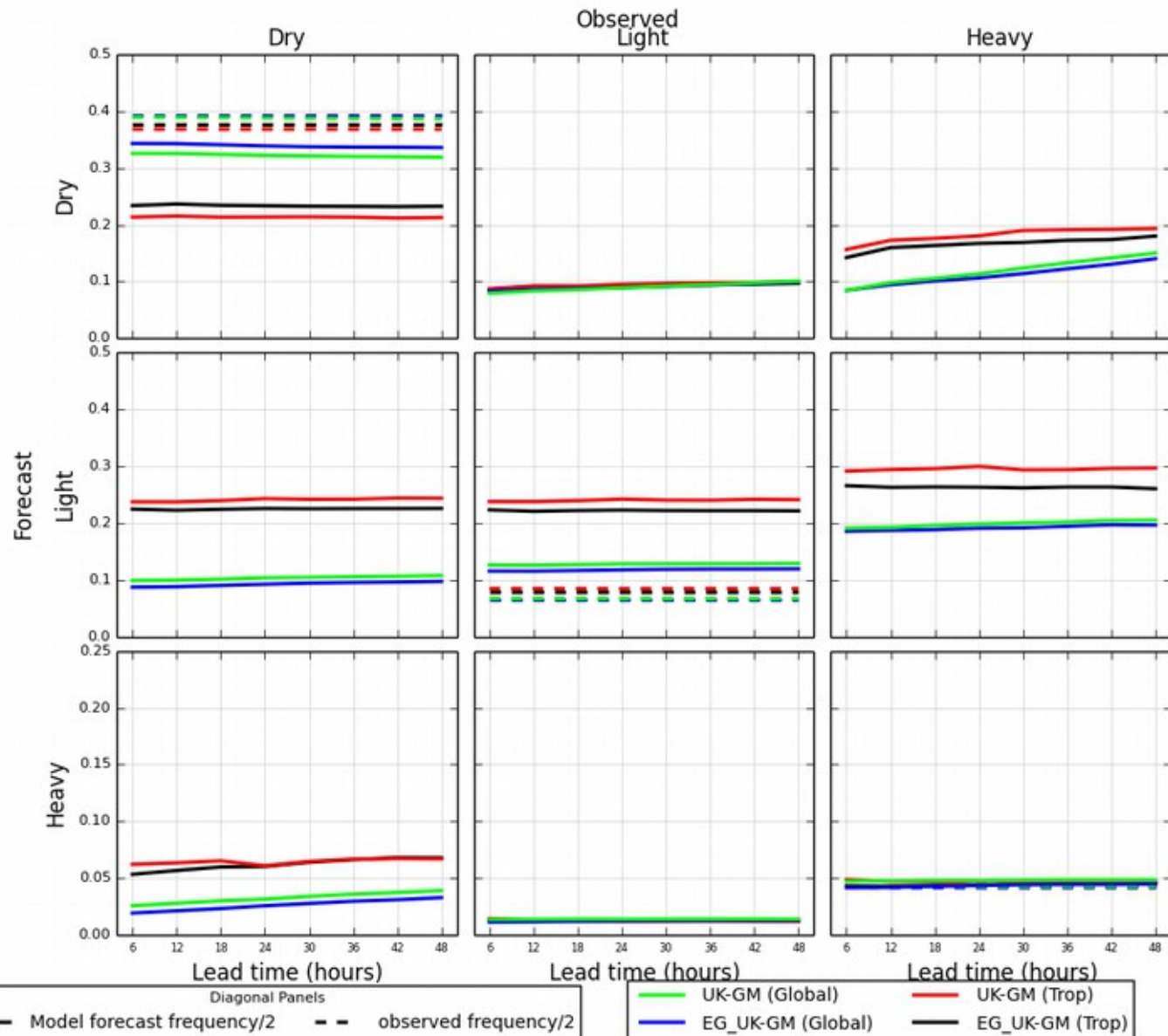
# Verification of QPF using SEEPS

## Decomposition into constituent error sources

Diurnal  
Average  
2012-2016

6-hour  
totals

UK-GM, EG\_UK-GM diurnal average SEEPS decomposition  $S_{ef}$  trial average over dates (20120401 to 20160229)



# Verification of QPF using SEEPS

## Decomposition into constituent error sources

- Sources contributing most to SEEPS score from 24-h accumulations are the **observed dry/forecast light** (especially in the Tropics) and **the observed heavy/forecast light** error categories. Both categories contribute similar proportions to the total score.
- Largest fraction of SEEPS score for 6-h accumulations is contributed by **observed heavy/forecast light**.
- Drop in skill in 24-h scores over Tropics is from **observed dry/forecast light** and **heavy** and the **observed heavy/forecast light** categories.
- Similar story for the 6-h scores over Tropics, but with the addition of the contribution from the **observed heavy/forecast dry**.
- Under-prediction of the number of dry events for both 6-h and 24-h accumulations.
- Over-prediction of the number of light precipitation events for both 6-h and 24-h totals.



# Verification of QPF using SEEPS

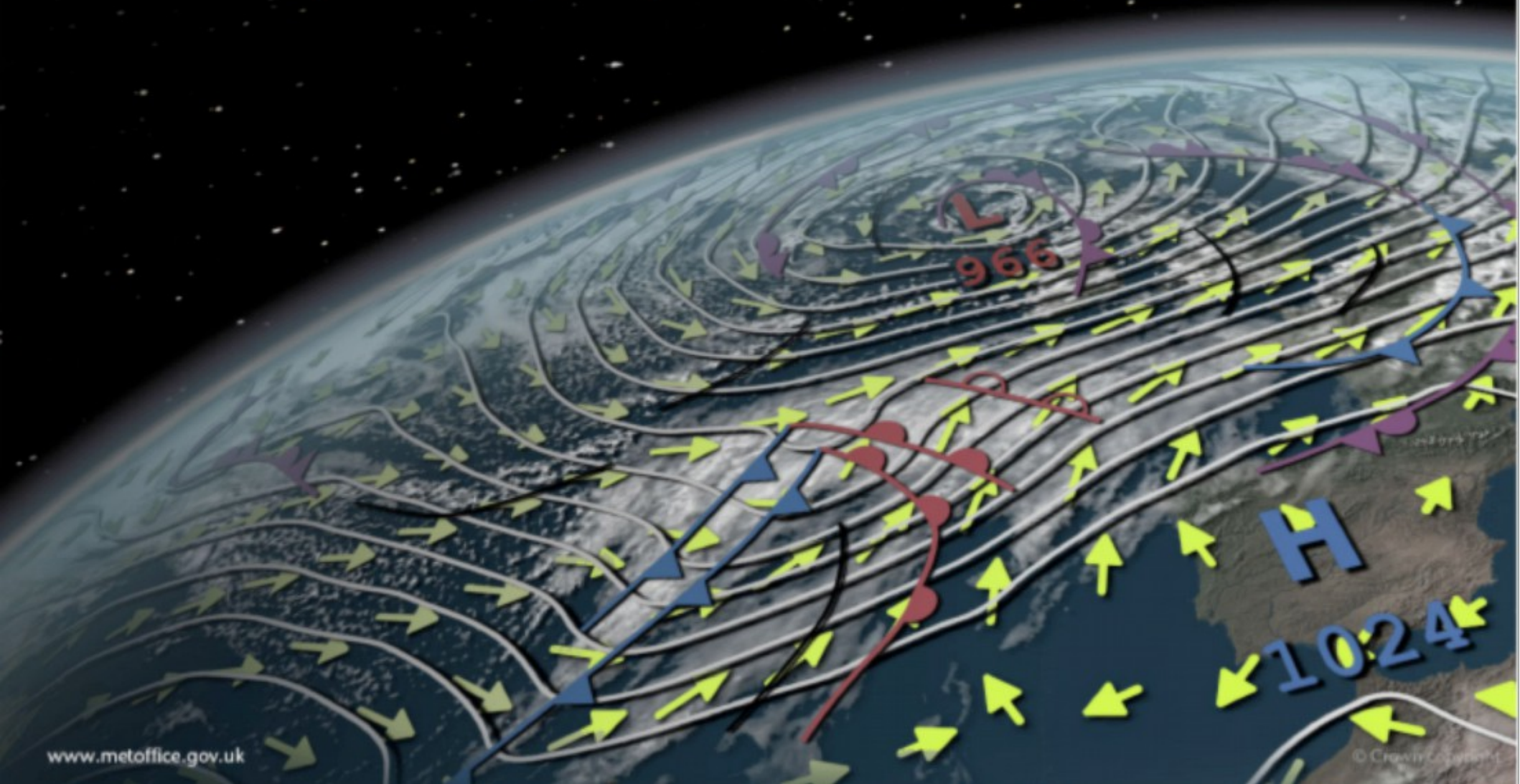
## Decomposition into constituent error sources

- UM GA6 still indicating improved frequency bias in number of dry events for both 6-hour and daily totals.
- Dips in skill seen in Northern Hemisphere summer (associated with convection and due to domination of Northern Hemisphere sites to the aggregated total score).
- Missed heavy events are penalised more at longer lead times, and a large source to error score.



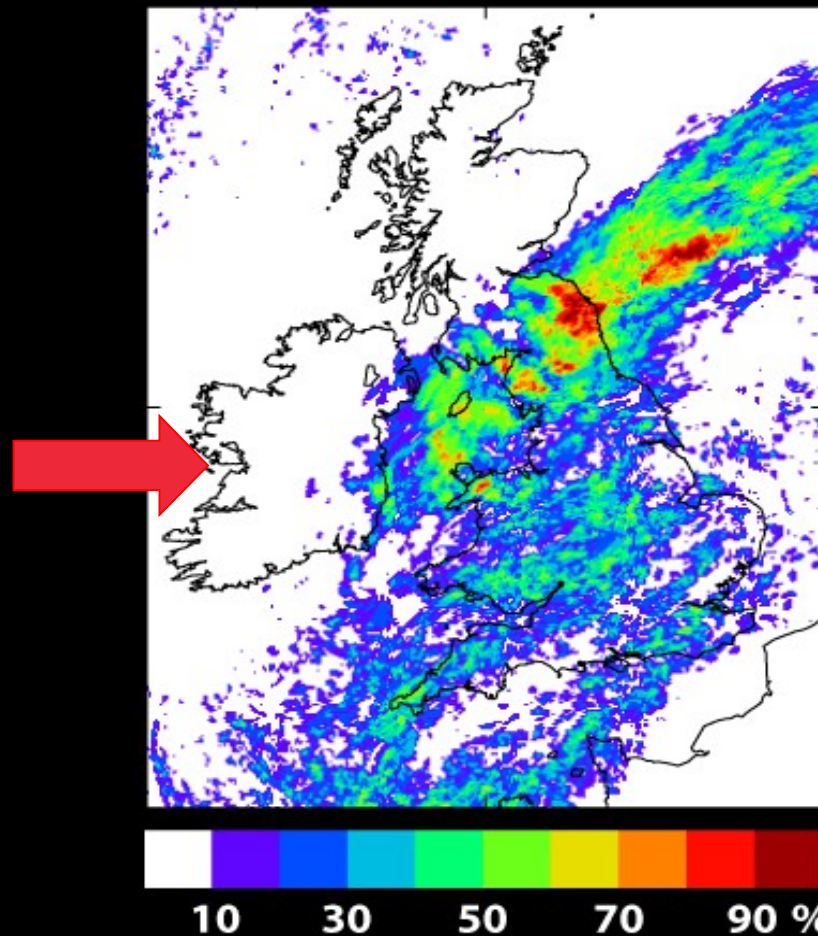
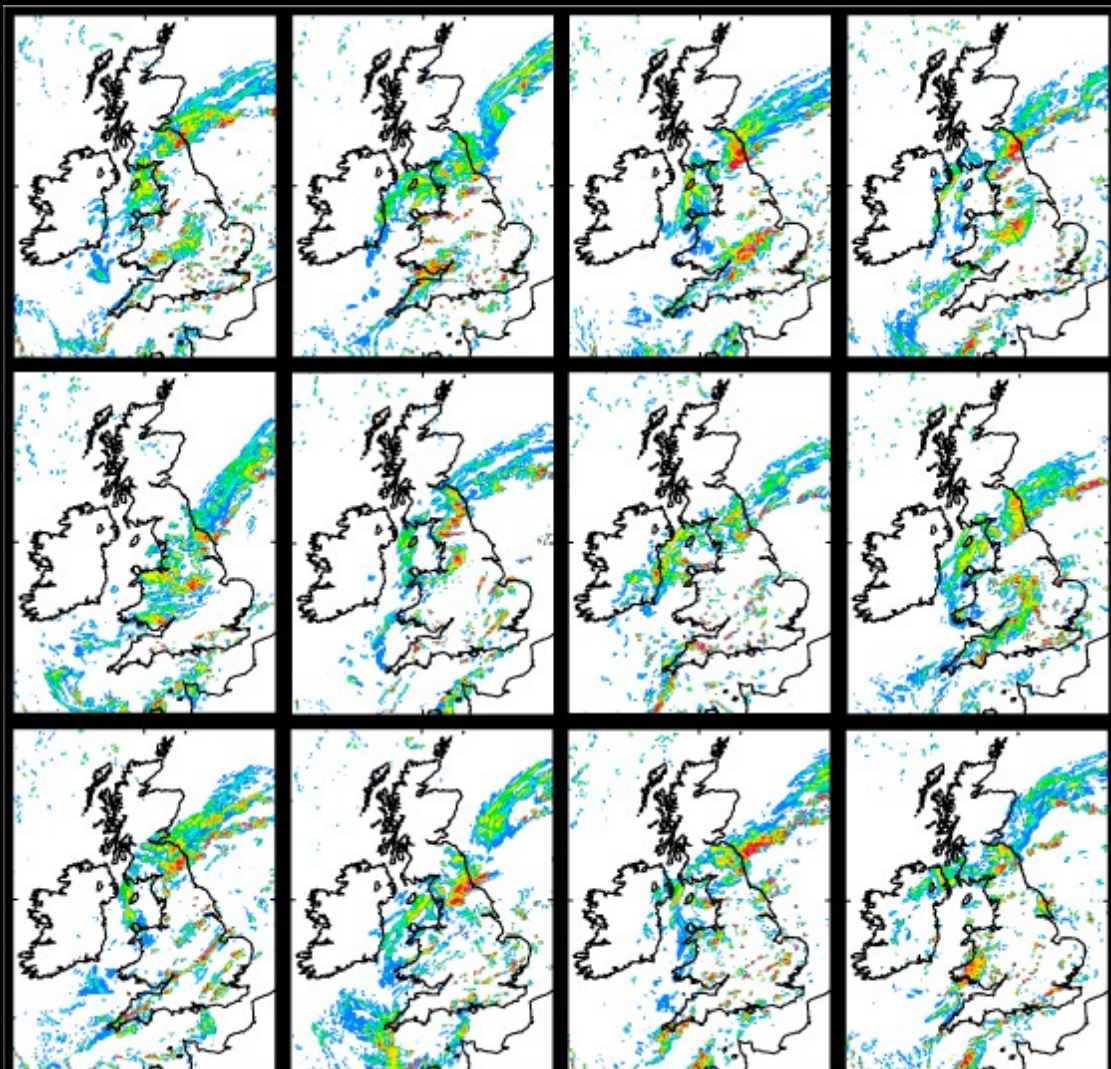
# Regional model precipitation verification

Researching use of neighbourhood processing for probabilistic prediction





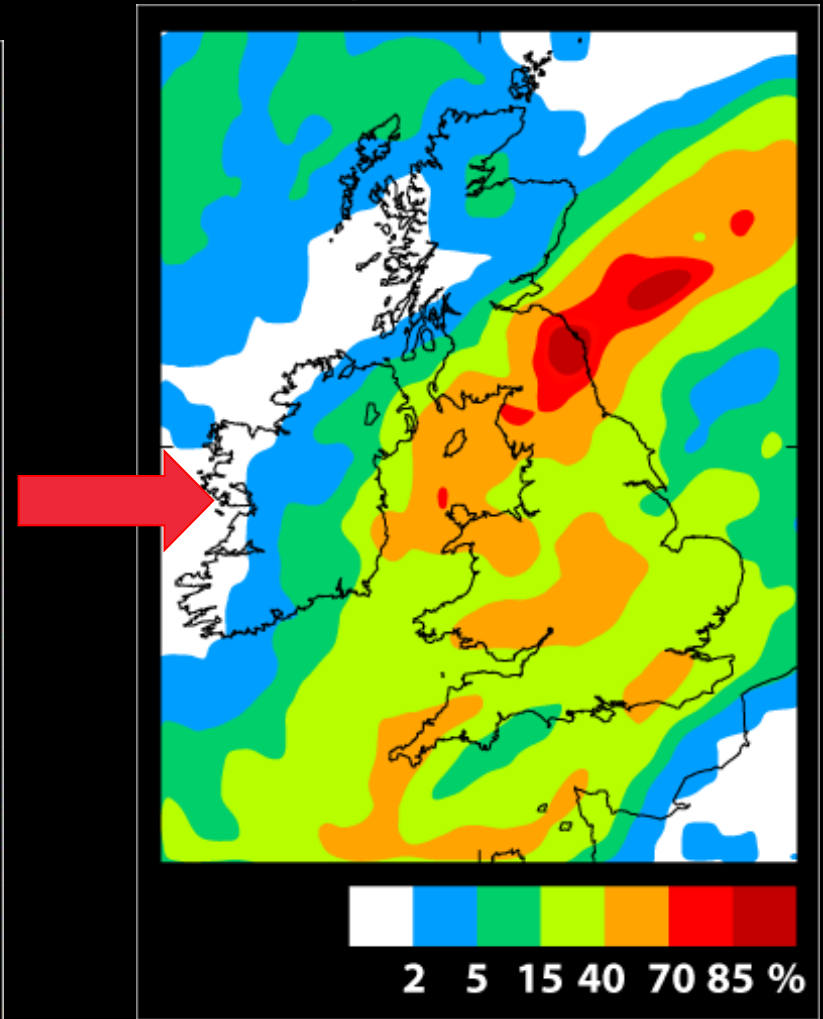
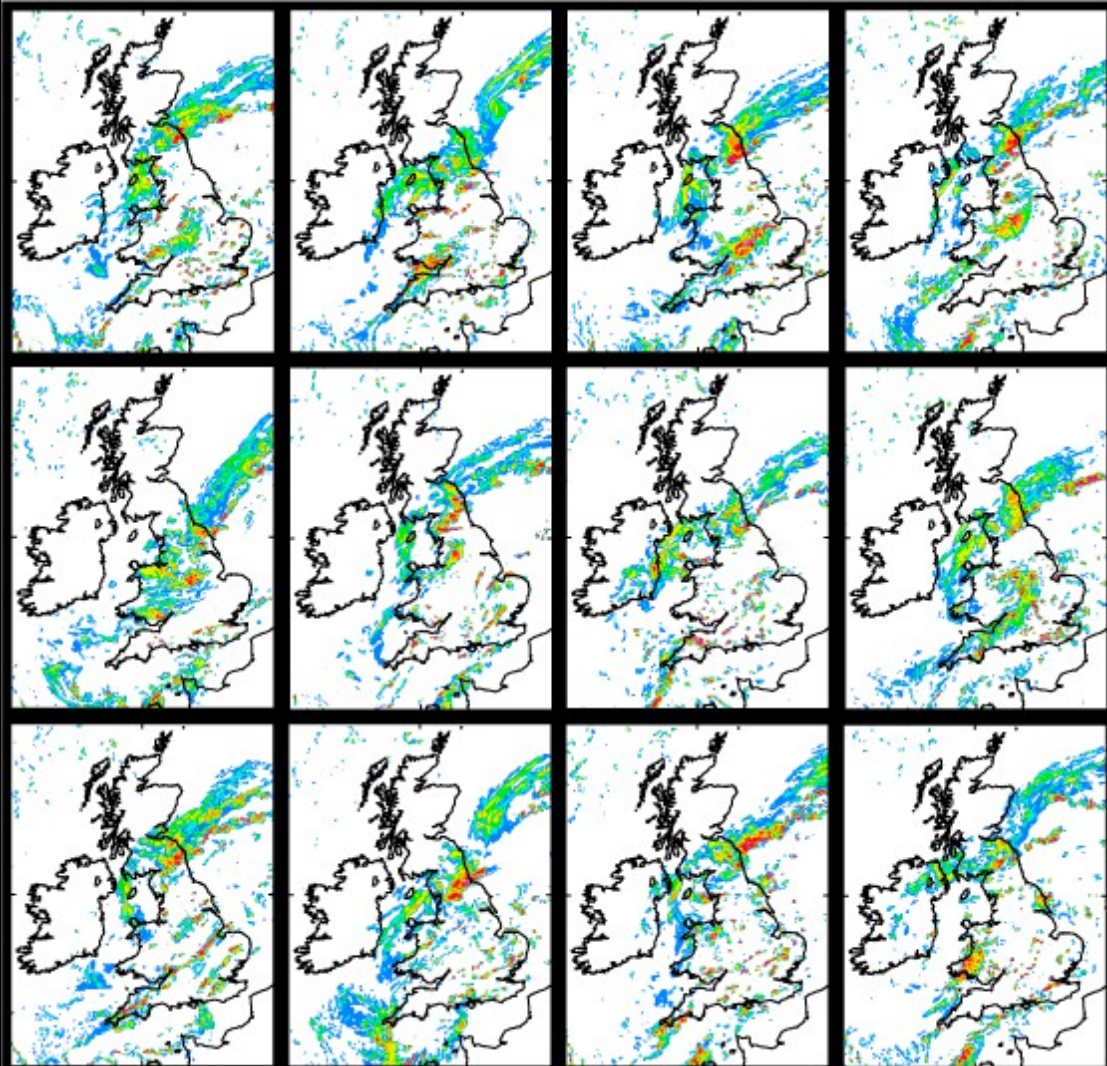
# MOGREPS-UK 2.2km ensemble



Undersampling leaves “holes” of zero-probability where showers could still occur



# MOGREPS-UK ... with Neighbourhood processing



Holes filled in

**NCEP**

# QPF Verification at NCEP for Deterministic NCEP and International Models

- 24h (12Z-12Z) contingency table-based verifications for all models
- 24h (12Z-12Z) contingency table-based verification for Alaska, Hawaii, Puerto Rico for select NCEP opnl and para models (observation/analysis quality not as high as over ConUS)
- 3-hourly contingency table-based verifications for NCEP operational and parallel models
- Contingency table-based scores include FB, POD, FAR, POFD, TS, ETS, HK, HSS, OR, EDI, SEDS, SEDI and many others
- 24h FSS computation for NCEP operational and parallel models
- 6-hourly FSS for select NCEP operational and parallel models, beginning Aug 2015
- International models verified over ConUS: CMC (global and regional), DWD, ECMWF, JMA, METFR, UKMO (24h only)



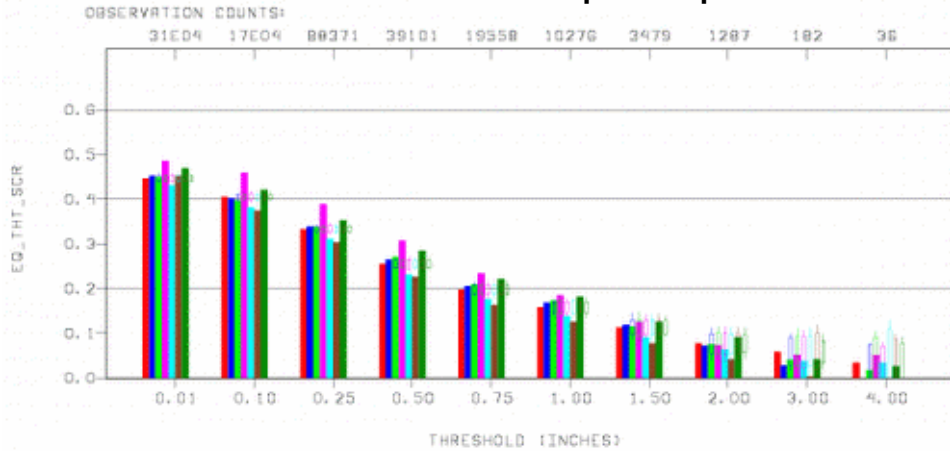
# ETS/Bias over ConUS, 1/2/3-day fcsts of Global Models

STAT=FHO PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=G211/RFC VYMDH=201504010000-201509302300 CI ALPHA=0.050

- MODEL=GFS
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=METFR
- MODEL=UKMO

BOX CONF INT = 0.950  
# OF SAMPLES = 2000

Apr-Sept 2015

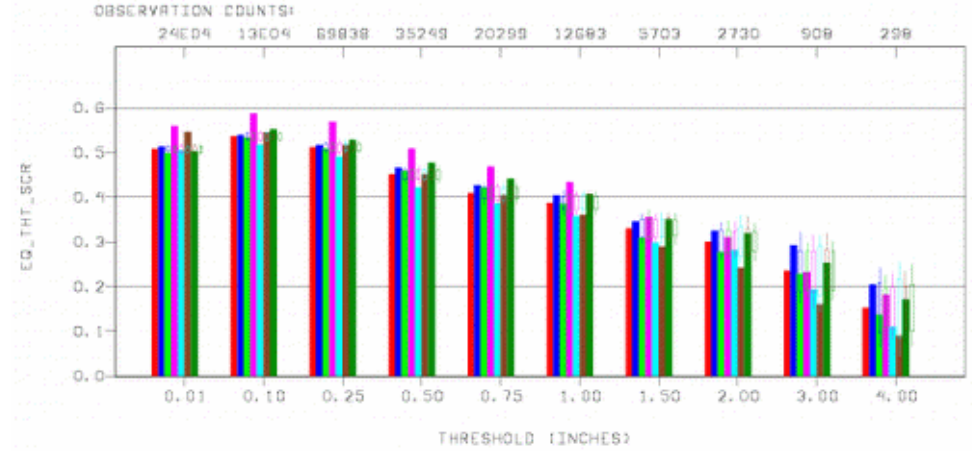


STAT=FHO PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=G211/RFC VYMDH=201510010000-201603312300 CI ALPHA=0.050

- MODEL=GFS
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=METFR
- MODEL=UKMO

BOX CONF INT = 0.950  
# OF SAMPLES = 2000

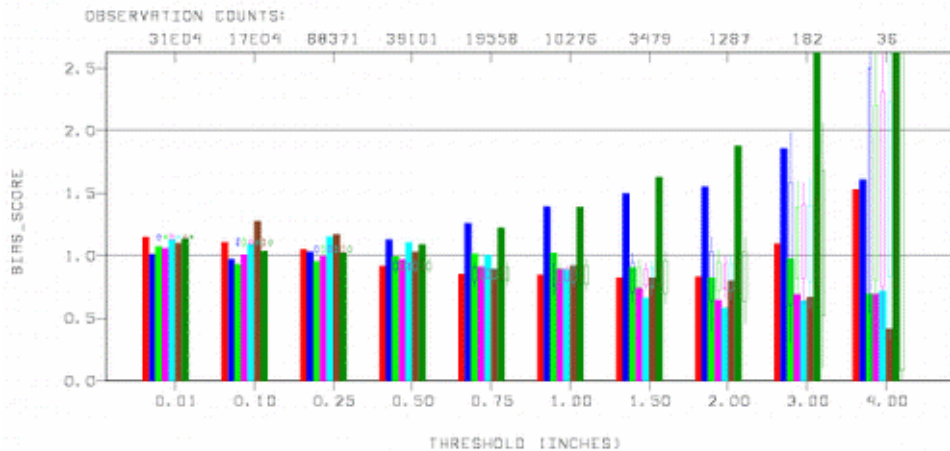
Oct 2015-Mar 2016



STAT=FHO PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=G211/RFC VYMDH=201504010000-201509302300 CI ALPHA=0.050

- MODEL=GFS
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=METFR
- MODEL=UKMO

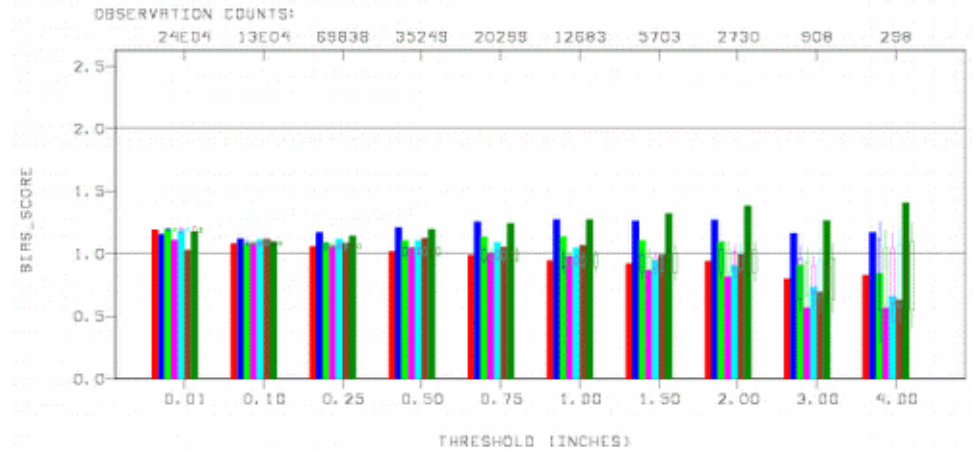
BOX CONF INT = 0.950  
# OF SAMPLES = 2000



STAT=FHO PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=G211/RFC VYMDH=201510010000-201603312300 CI ALPHA=0.050

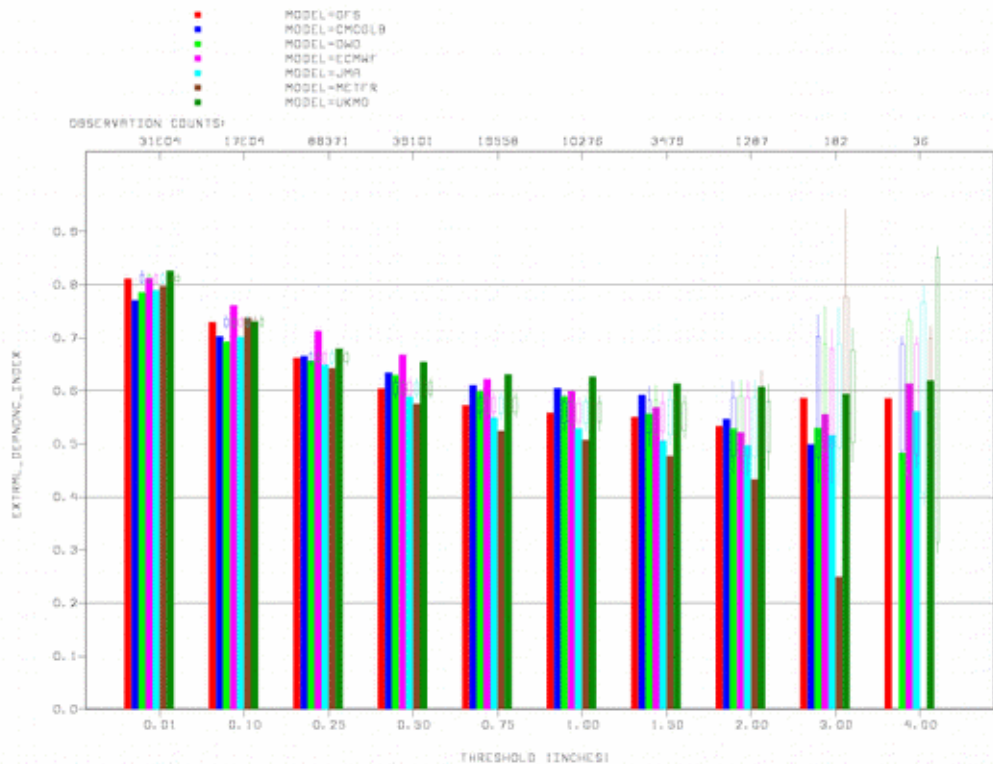
- MODEL=GFS
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=METFR
- MODEL=UKMO

BOX CONF INT = 0.950  
# OF SAMPLES = 2000

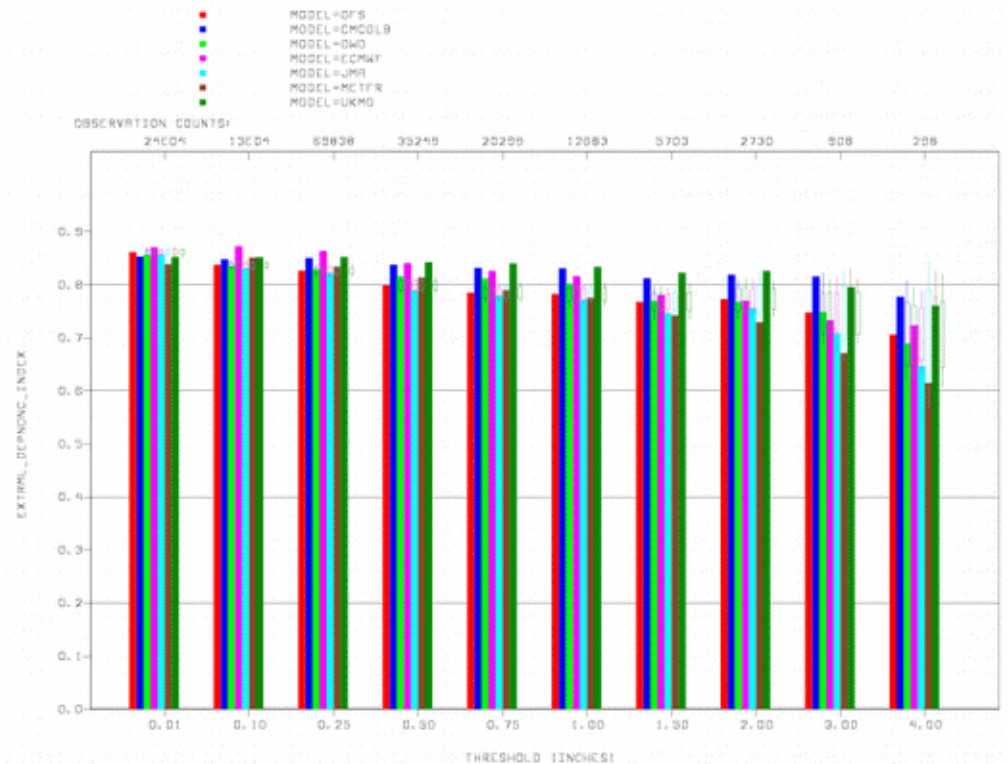


# Extremal dependence index over ConUS 1/2/3-day fcsts of Global Models

STAT=FHO PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=G211/RFC VYMDH=201504010000-201509302300 CI ALPHA=0.051 STAT=FHO PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=G211/RFC VYMDH=201510010000-201603312300 CI ALPHA=0.051



Apr-Sept 2015



Oct 2015-Mar 2016

# ETS/Bias over ConUS, 1/2-day fcsts of Global Models and NAM/CMC (CMC fcst to 48h)

STAT=FHO PARAM=APCP/24 FHOURL=24+48 V\_RGN=6211/RFC VYMDH=201504010000-201509302300 CI ALPHA=0.050

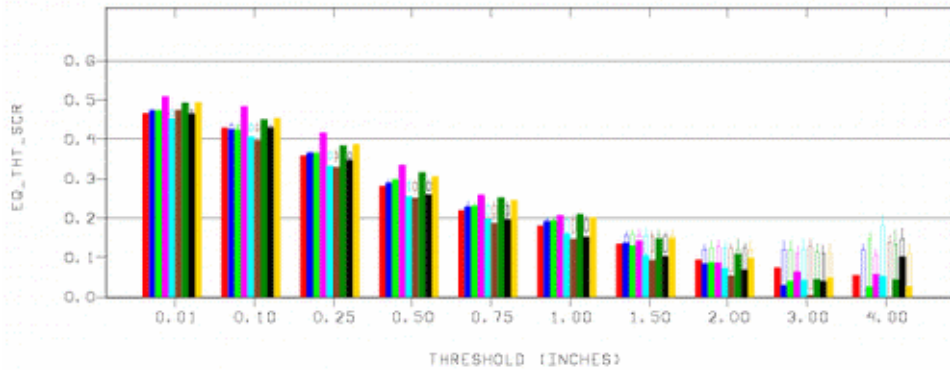
- MODEL=GF5
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=METFR
- MODEL=UKMO
- MODEL=NAM
- MODEL=CMC

BOX CONF INT = 0.950  
# OF SAMPLES = 2000

Apr-Sept 2015

OBSERVATION COUNTS:

21ED4 11E04 58963 26102 13048 6859 2318 855 124 24



STAT=FHO PARAM=APCP/24 FHOURL=24+48 V\_RGN=6211/RFC VYMDH=201510010000-201603312300 CI ALPHA=0.050

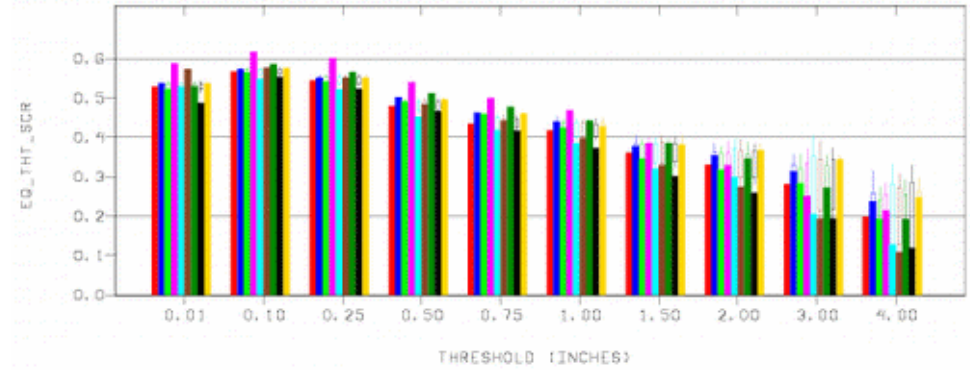
- MODEL=GF5
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=METFR
- MODEL=UKMO
- MODEL=NAM
- MODEL=CMC

BOX CONF INT = 0.950  
# OF SAMPLES = 2000

Oct 2015-Mar 2016

OBSERVATION COUNTS:

16ED4 83325 46459 23411 13440 8379 3776 1826 605 196



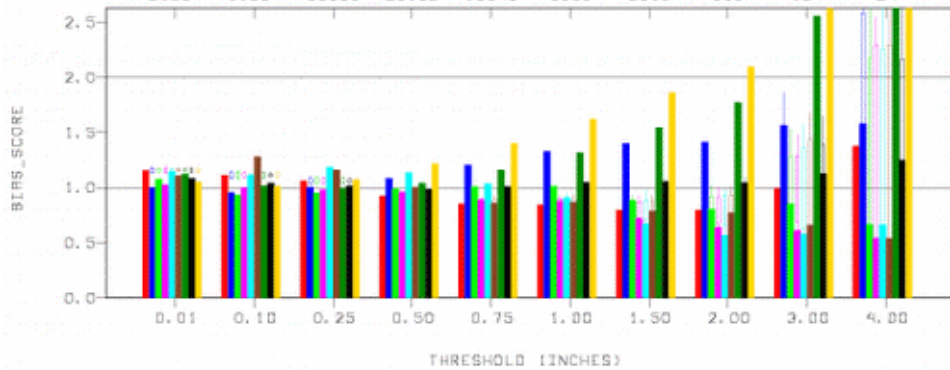
STAT=FHO PARAM=APCP/24 FHOURL=24+48 V\_RGN=6211/RFC VYMDH=201504010000-201509302300 CI ALPHA=0.050

- MODEL=GF5
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=METFR
- MODEL=UKMO
- MODEL=NAM
- MODEL=CMC

BOX CONF INT = 0.950  
# OF SAMPLES = 2000

OBSERVATION COUNTS:

21ED4 11E04 58963 26102 13048 6859 2318 855 124 24



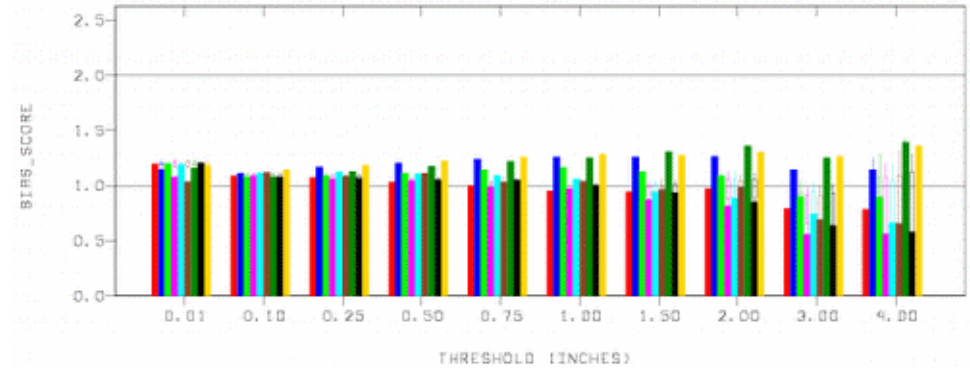
STAT=FHO PARAM=APCP/24 FHOURL=24+48 V\_RGN=6211/RFC VYMDH=201510010000-201603312300 CI ALPHA=0.050

- MODEL=GF5
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=METFR
- MODEL=UKMO
- MODEL=NAM
- MODEL=CMC

BOX CONF INT = 0.950  
# OF SAMPLES = 2000

OBSERVATION COUNTS:

16ED4 83325 46459 23411 13440 8379 3776 1826 605 196



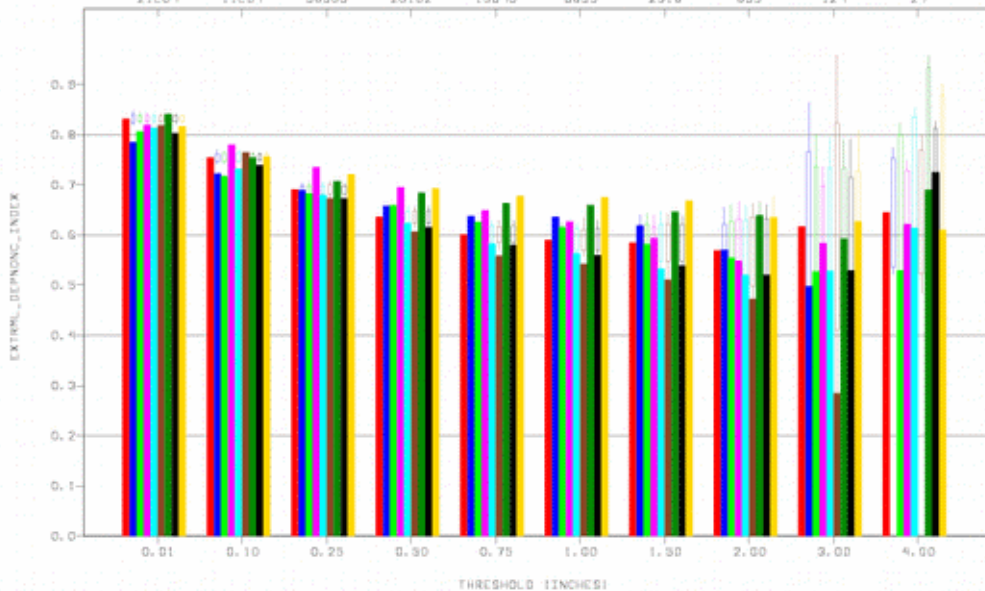
# Extremal Dependence index over ConUS, 1/2-day fcsts of Global Models and NAM/CMC (CMC fcst to 48h)

STAT=FHD PARAM=RPCP/24 FHOUR=24+48 V\_RGN=G211/RFC VYMOH=201504010000-201509302300 CI ALPHA=0.050

- MODEL=DFS
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=MC TFR
- MODEL=UKMO
- MODEL=NAM
- MODEL=CMC

OBSERVATION COUNTS:

21004 11004 50863 26102 13048 6859 2318 855 124 24



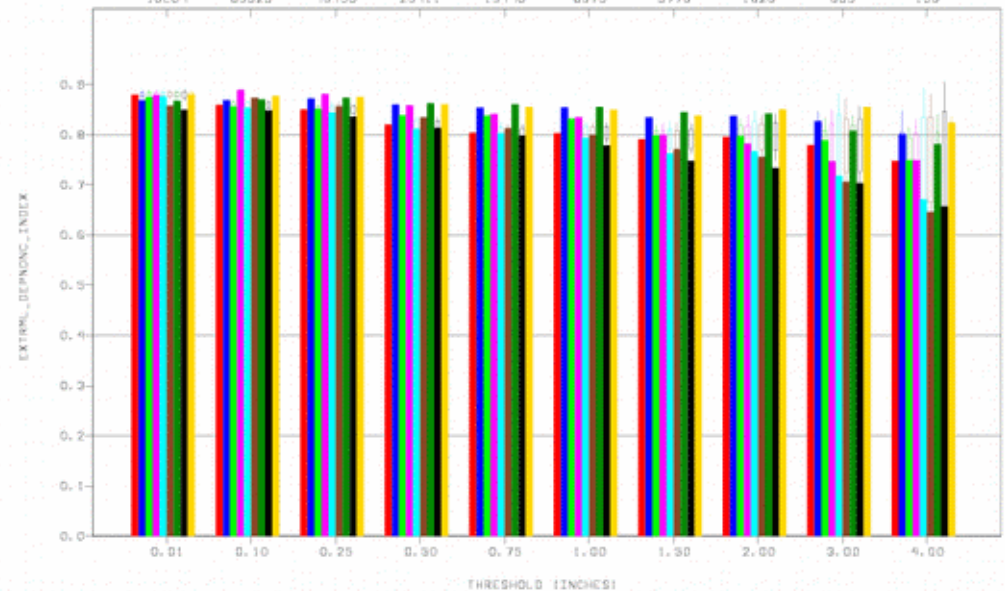
Apr-Sept 2015

STAT=FHD PARAM=RPCP/24 FHOUR=24+48 V\_RGN=G211/RFC VYMOH=201510010000-201603312300 CI ALPHA=0.050

- MODEL=DFS
- MODEL=CMCGLB
- MODEL=DWD
- MODEL=ECMWF
- MODEL=JMA
- MODEL=MC TFR
- MODEL=UKMO
- MODEL=NAM
- MODEL=CMC

OBSERVATION COUNTS:

18004 83325 46458 23411 13440 8378 3775 1825 605 195

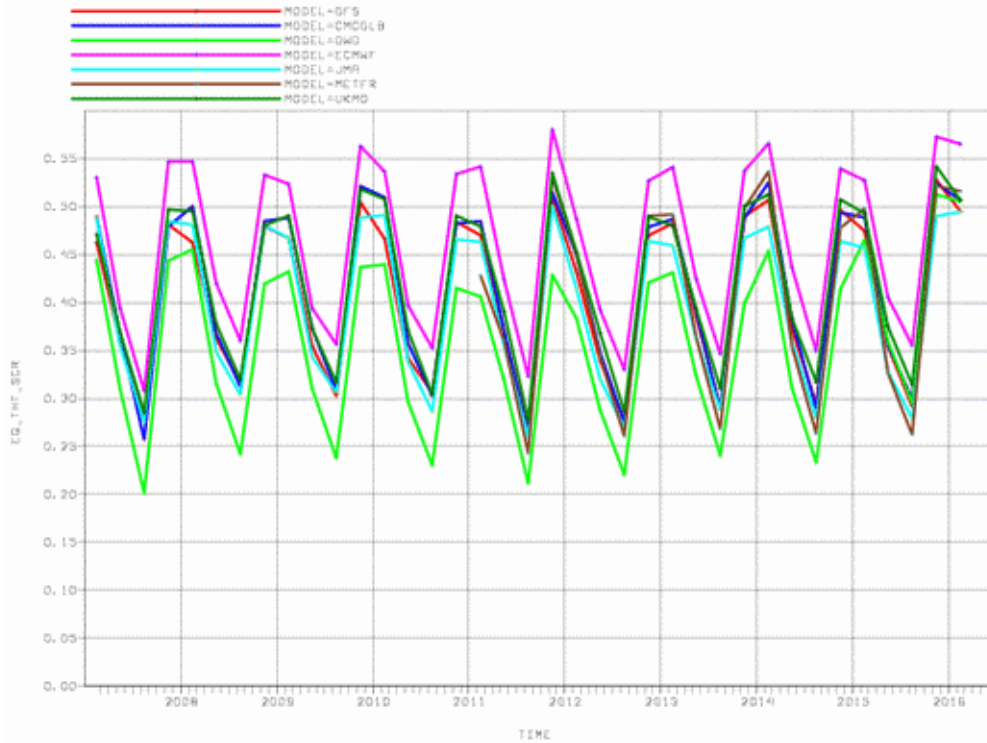


Oct 2015-Mar 2016



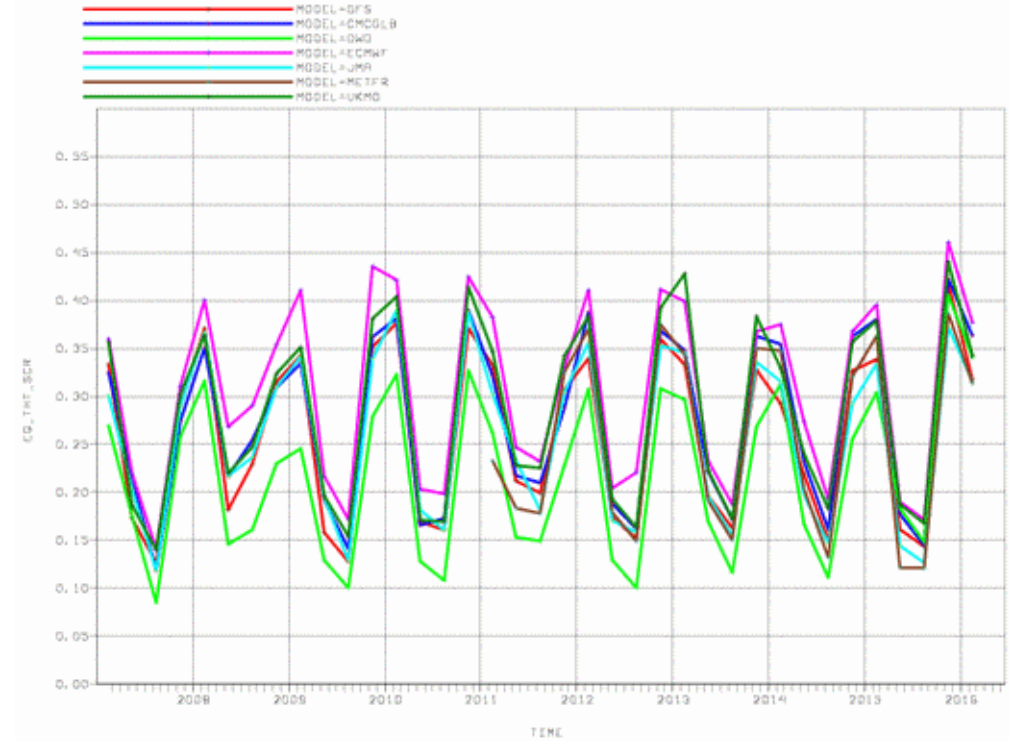
# Quarterly time series of ETS, all global models

STAT=FHD PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=6211/RFC LEVEL=SFC THRS=0.25 VYMDH=200701010000-201603312300



6.35mm/day threshold

STAT=FHD PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=6211/RFC LEVEL=SFC THRS=1.00 VYMDH=200701010000-201603312300

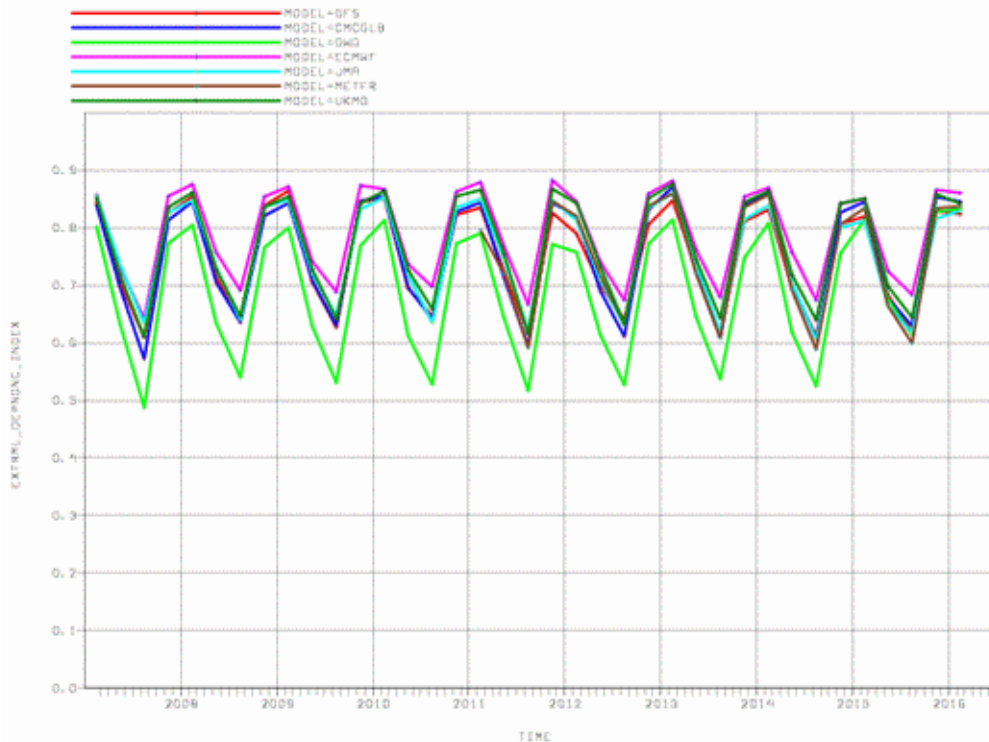


25.4mm/day threshold

(verification for Metéo-France began in Mar 2011)

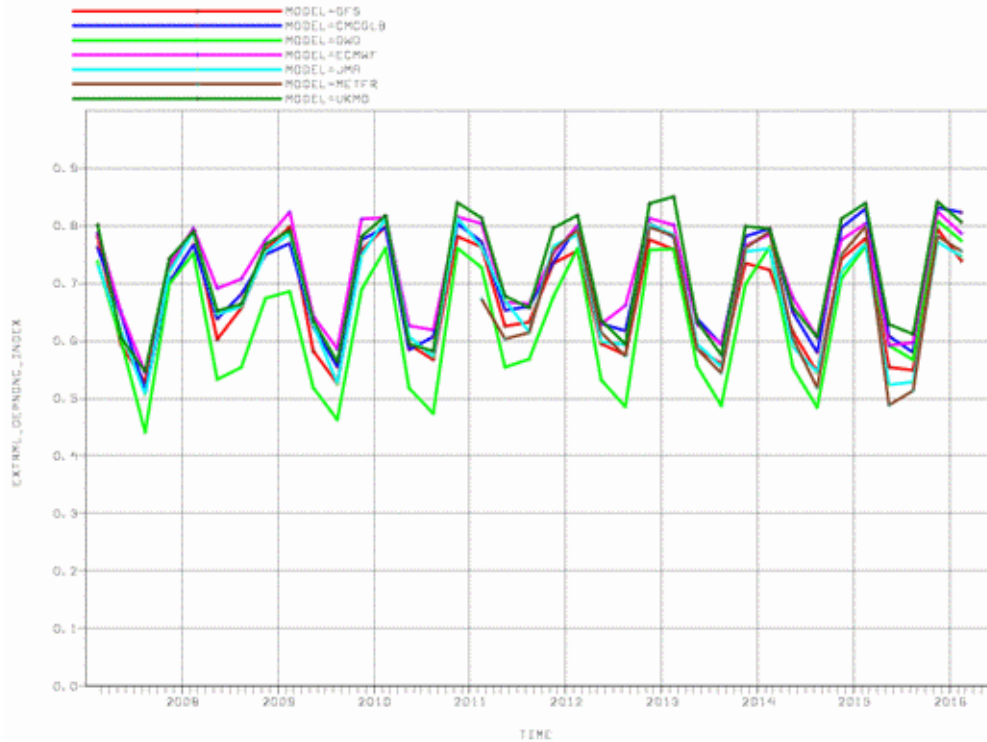
# Quarterly time series of Extremal Dependence Index all global models

STAT=FHD PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=6211/RFC LEVEL=SFC THRS=0.25 VYMDH=200701010000-201603312300



6.35mm/day threshold

STAT=FHD PARAM=APCP/24 FHOUR=24+48+72 V\_RGN=6211/RFC LEVEL=SFC THRS=1.00 VYMDH=200701010000-201603312300



25.4mm/day threshold

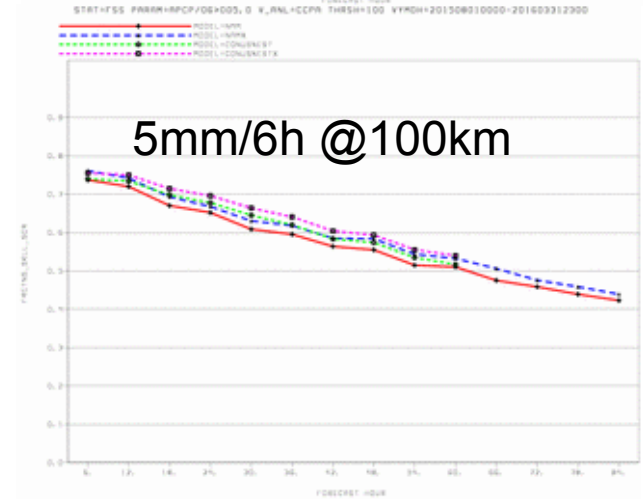
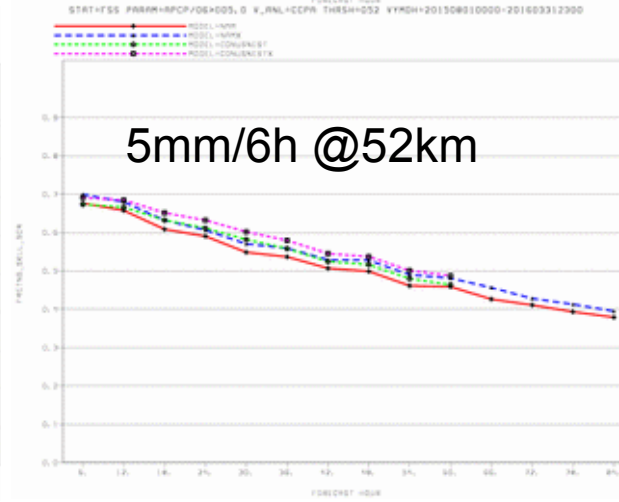
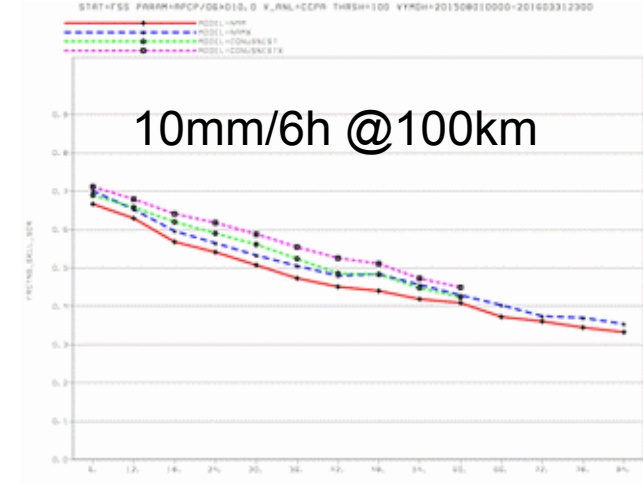
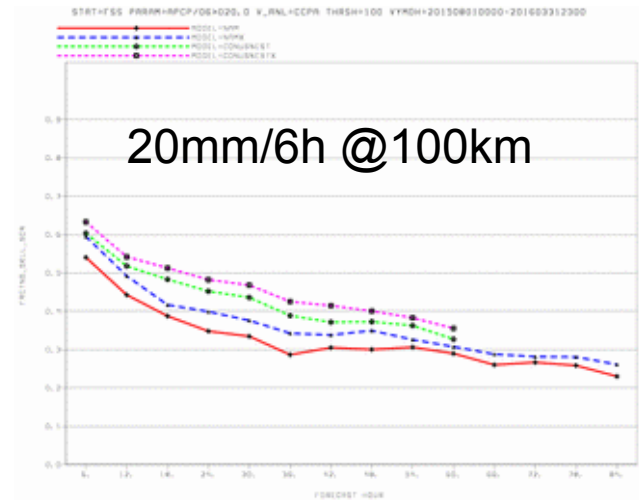
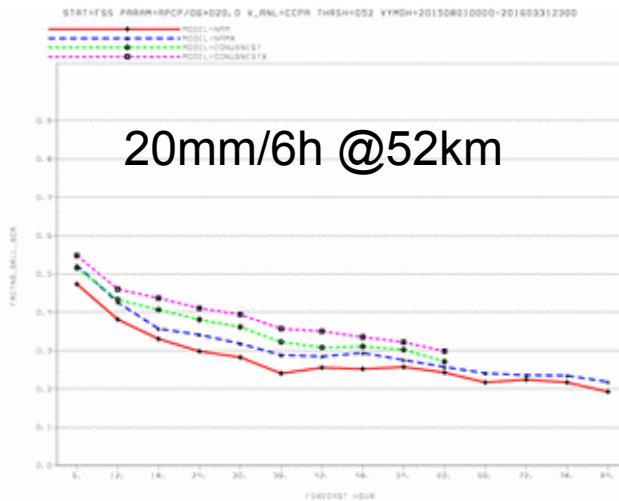
(verification for Metéo-France began in Mar 2011)



## Moving toward adopting more of the JWGFVR recommendations

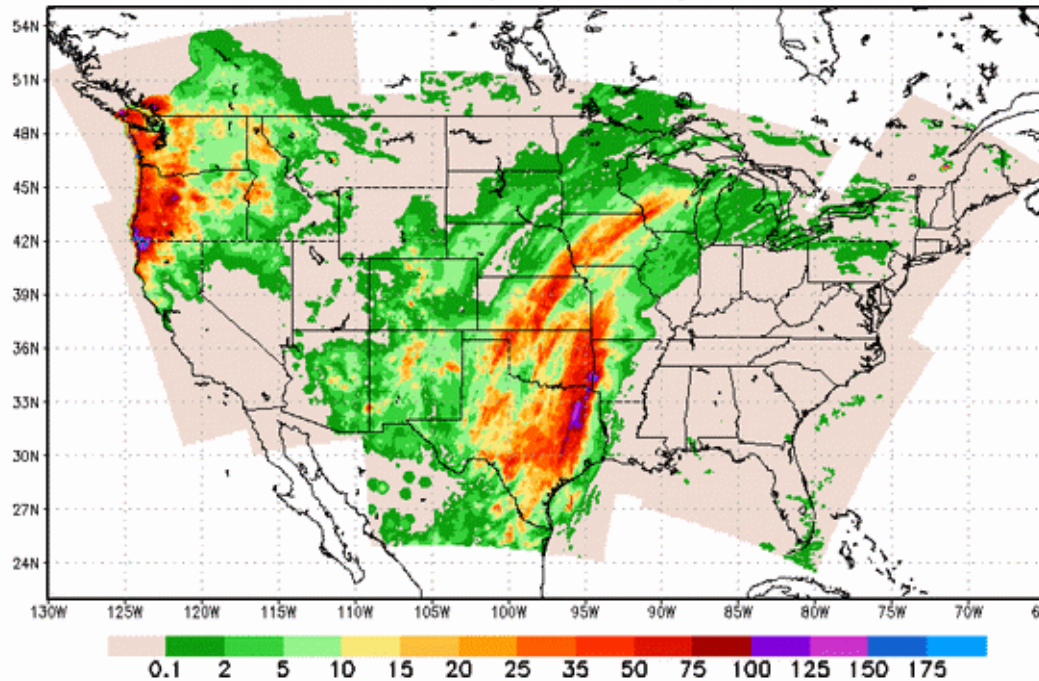
- Began routine 6-hourly FSS calculations for NCEP models (using recommended thresholds in SI units). Examples show on P9. Would like more international model 6-hourly forecast files made available.
- Began verification on nearest gauge locations (using recommended thresholds in SI units)

# NAM/NAMX/CONUSNEST/CONUSNESTX Aug 2015 – Mar 2016, 6h FSS



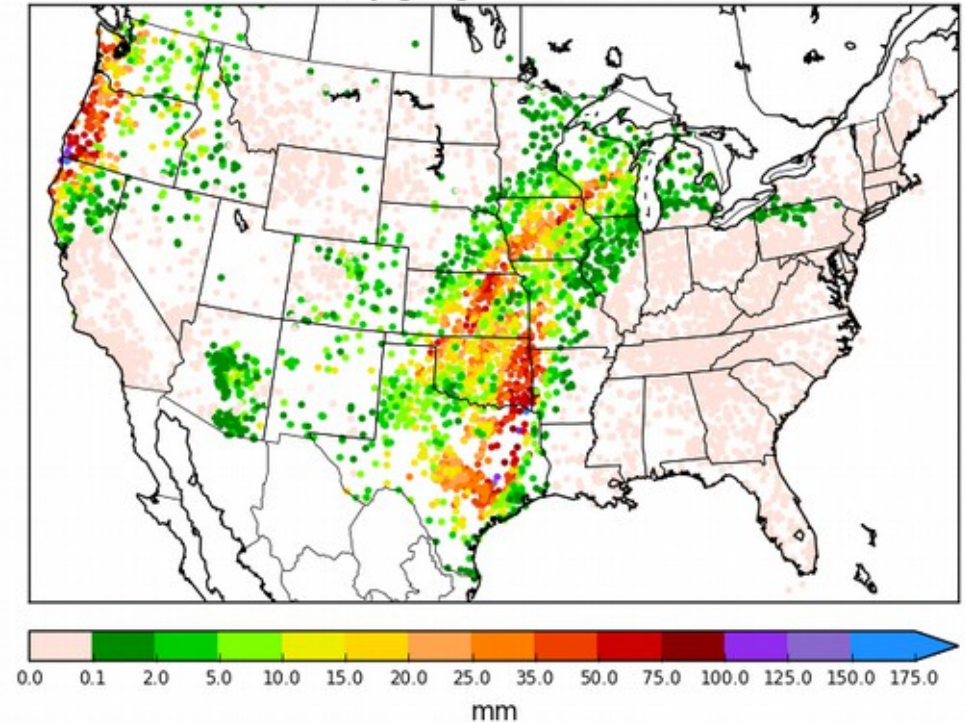
# Verifying analysis/gauges

CCPA 24h Accum (mm) Ending 2015121312



CCPA: NCEP Stage IV (radar+gauges) analysis with climate calibration. Used for 24h/6h/3h gridded verifications.

QC'd daily gauges, 12Z 20151213



~8,000 QC'd daily gauge reports.

# Additional Information

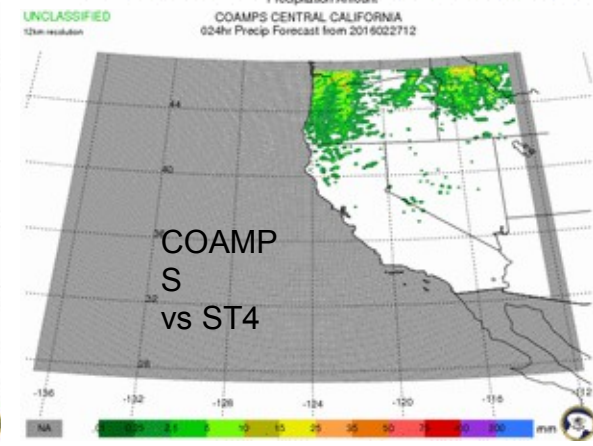
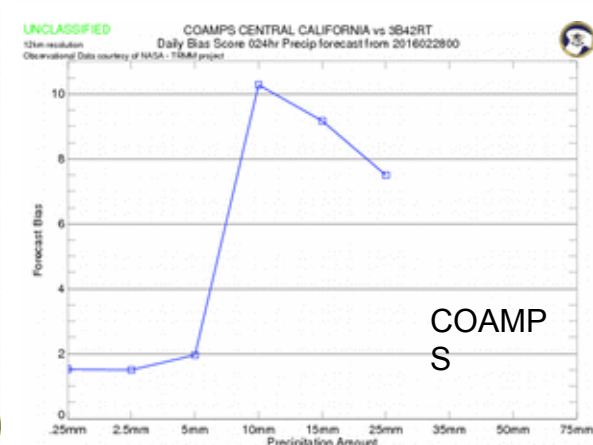
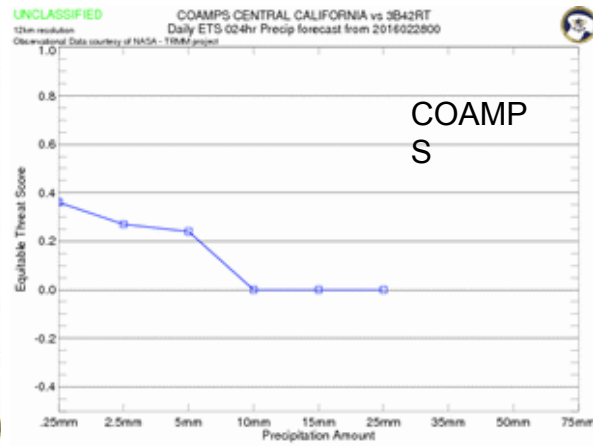
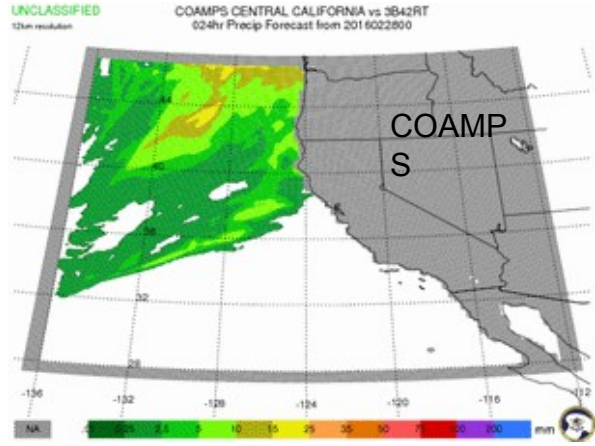
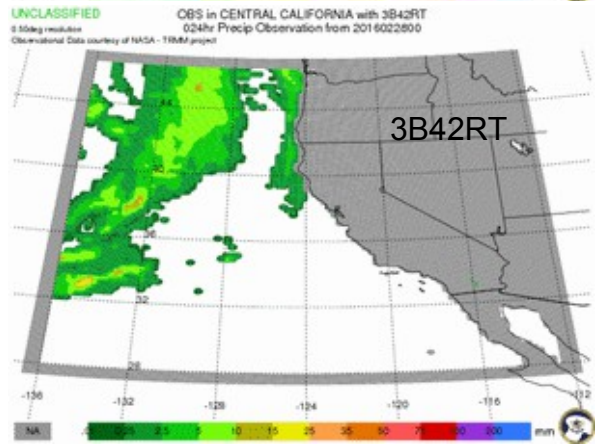
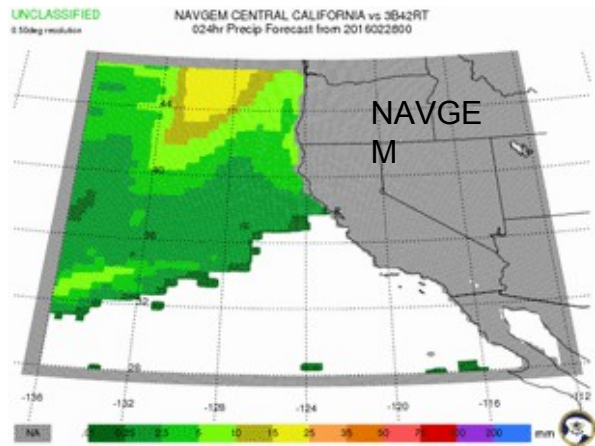
- GFS data made available to NCEP's international partners:  
<http://nomads.ncep.noaa.gov/pub/data1/nccf/com/verf/prod/precip.yyyymmdd/>
- Monthly precipitation scores of operational models:  
<http://www.emc.ncep.noaa.gov/mmb/ylin/pcpverif/scores/>
- Monthly precipitation scores of regional experimental model runs:  
<http://www.emc.ncep.noaa.gov/mmb/ylin/pcpverif/scores.paramodels/>
- Global experimental verification scores:  
[http://www.emc.ncep.noaa.gov/gmb/STATS\\_vsdb/](http://www.emc.ncep.noaa.gov/gmb/STATS_vsdb/)
- Daily side-by-side precipitation verification comparisons:  
<http://www.emc.ncep.noaa.gov/mmb/ylin/pcpverif/daily/>

**NRL**





# US Navy FNMOC QPF Verification

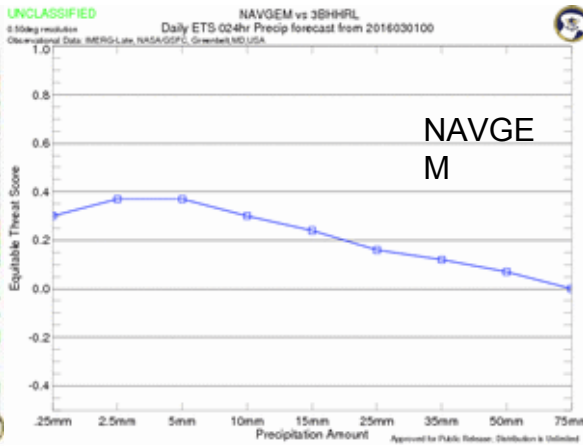
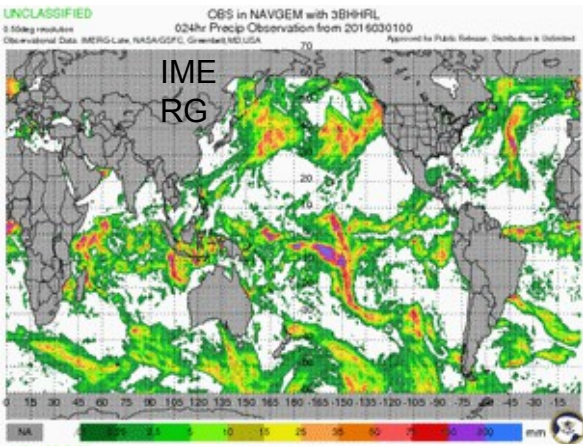


- Global and ~25 regionally relocatable regions verified operationally.
  - Over-Water verification against TRMM 3B42RT (soon IMERG 3BHHRL)
  - Over-Land verification against US NCEP ST4 and Aust. BoM radar/rain gauge sets
  - Bias, ETS, and HK at 9 thresholds (Trace to 75mm)
  - POD, POFD, FSS and other scores capable (not operationally done)
  - Monthly verification can be performed (not done operationally)
  - Code is NRLMRY and FNMOC collaboration
  - Global verification results publicly available online at FNMOC public verification page:  
[https://www.fnmoc.navy.mil/verify\\_](https://www.fnmoc.navy.mil/verify.cgi/)  
[cgi/](https://www.fnmoc.navy.mil/verify.cgi/)
- R. Lee (FNMOC), J. Nachamkin, T. Whitcomb, B. Ruston (NRL)

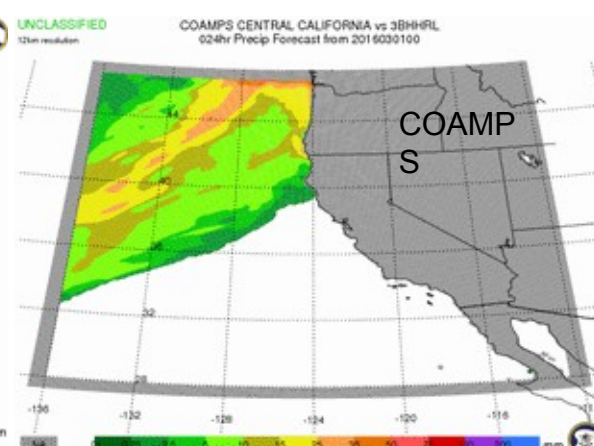
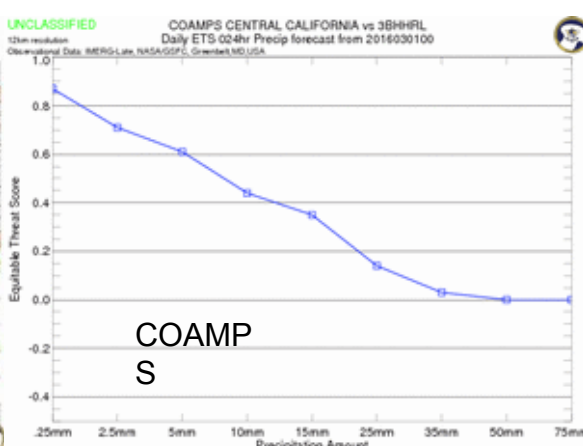
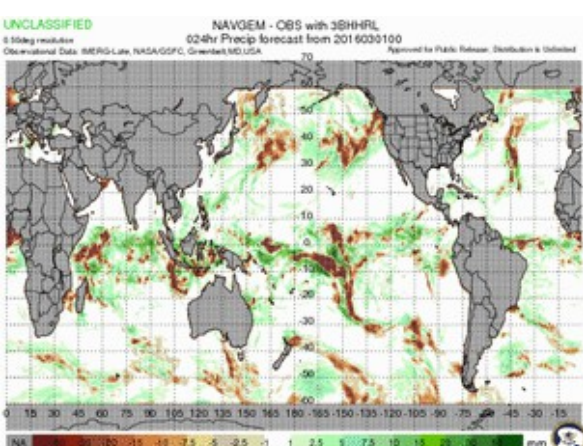
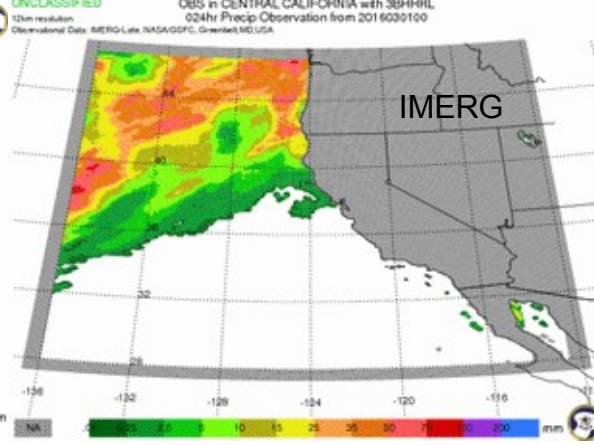
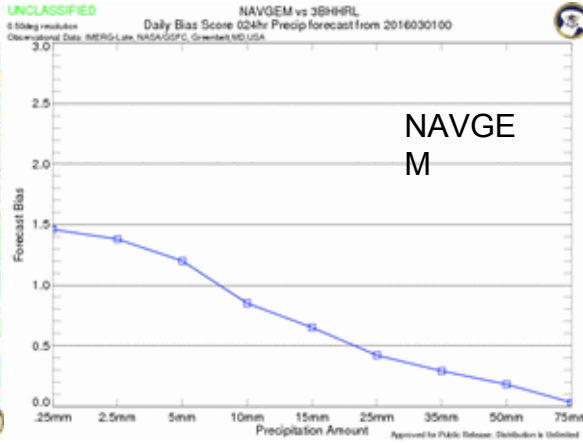
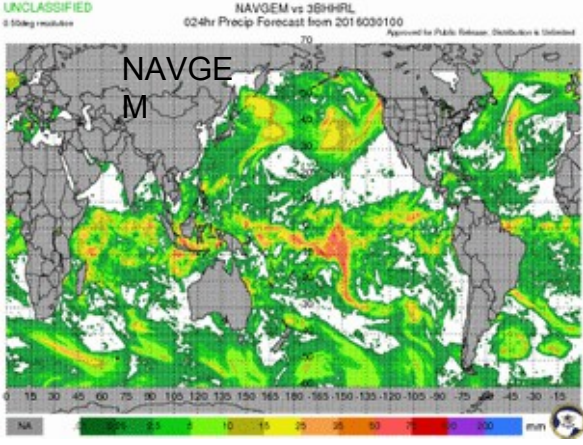




# QPF Verification Against NASA GSFC GPM/IMERG



Observational data from:  
G. Huffman, D. Bolvin, D. Braithwaite, K. Hsu, R. Joyce, P. Xie, 2014: Integrated Multi-satellitE Retrievals for GPM (IMERG), version V03E. NASA's Precipitation Processing Center, accessed 2016, <ftp://jsimpson.pps.eosdis.nasa.gov>



**RHMC**

Russian experiments on QPF  
assessment using high-  
resolution models

# Introduction

With gridded precipitation data from radar measurements and denser AMS networks becoming available, a number of recently developed verification methods and scores were introduced into practice at the Hydrometcentre of Russia of Roshydromet, however they are not yet operational:

scores for rare events, such as EDI

spatial verification methods

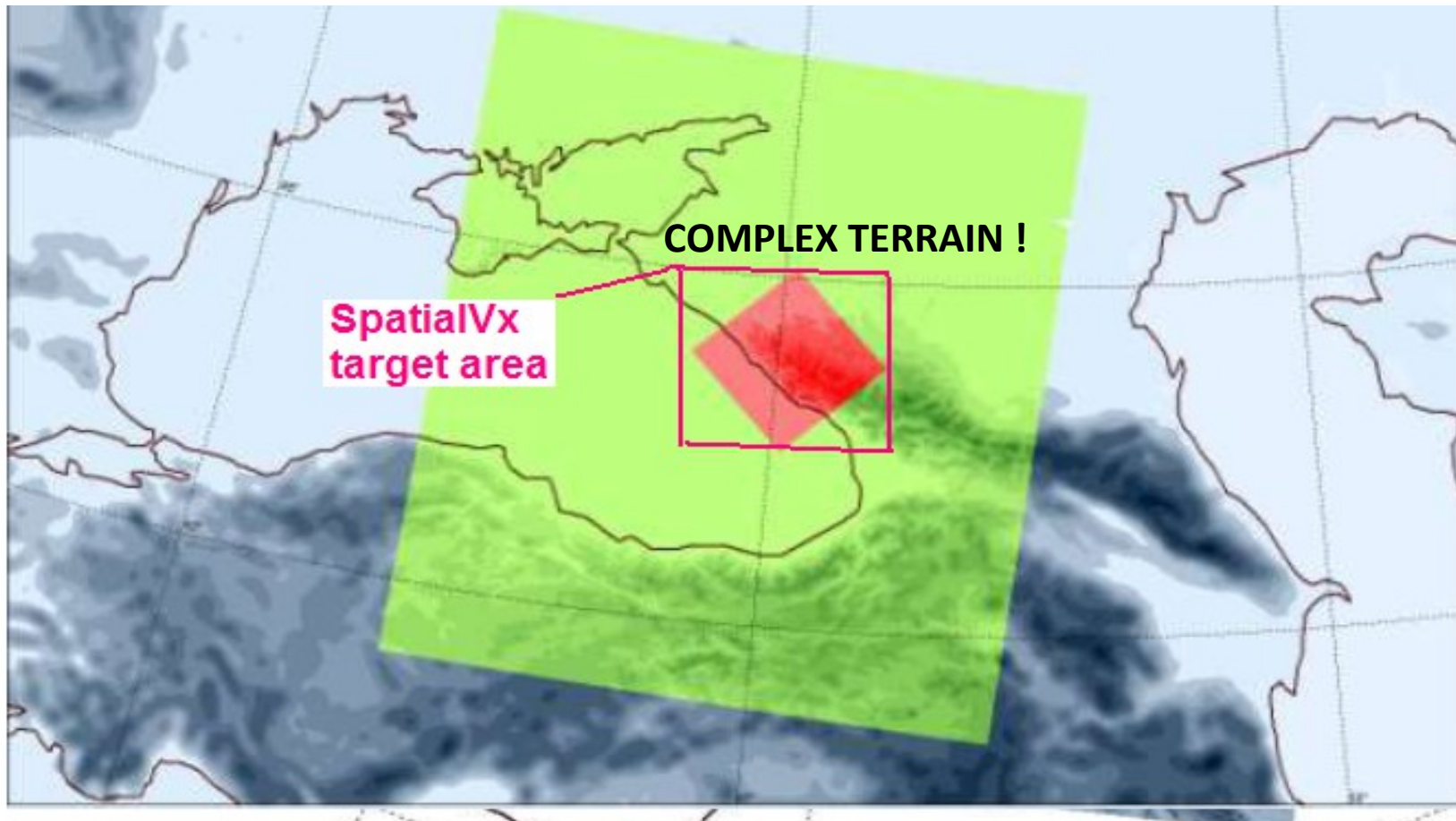
Many of these methods were tested in the Sochi area, as an extensive dataset was accumulated there during the Sochi-2014 Winter Olympic and Paralympic Games

# Area of the study

349 lon points \* 481 lat points with **0.00833** lat-lon increments.

1 grid size by **longitude** =  $111 * 0.00833 = 930$  m,

1 grid size by **latitude** =  $\cos(43^\circ 35') * 930$  m =  $0.72 * 930 = \sim 670$  m

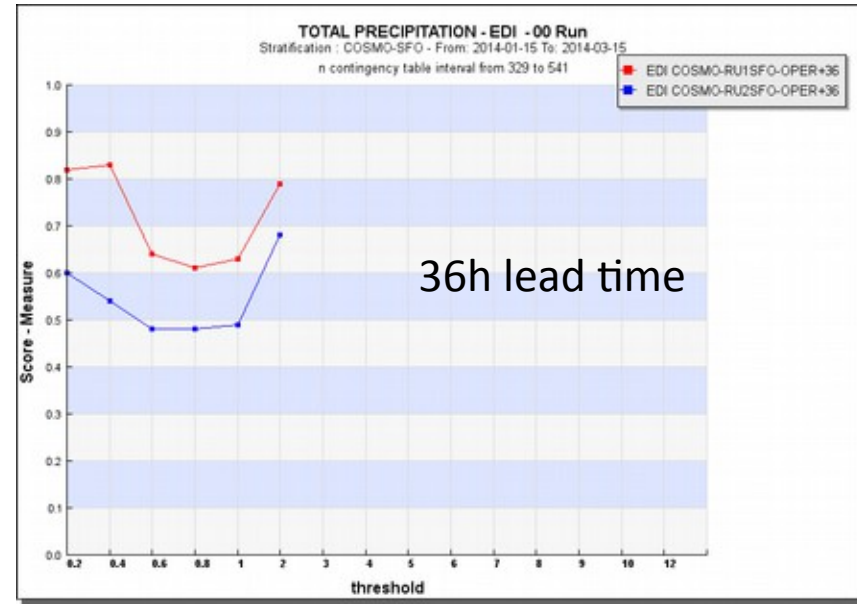
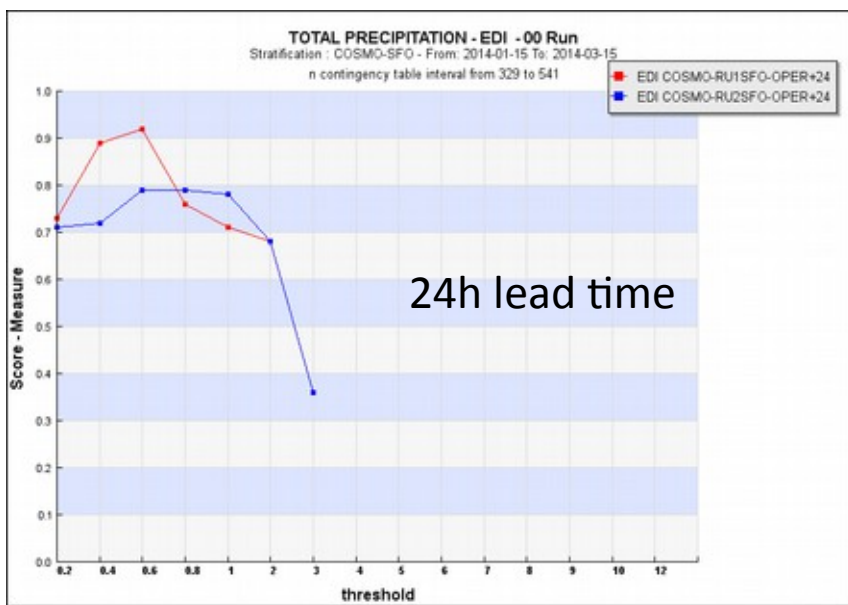
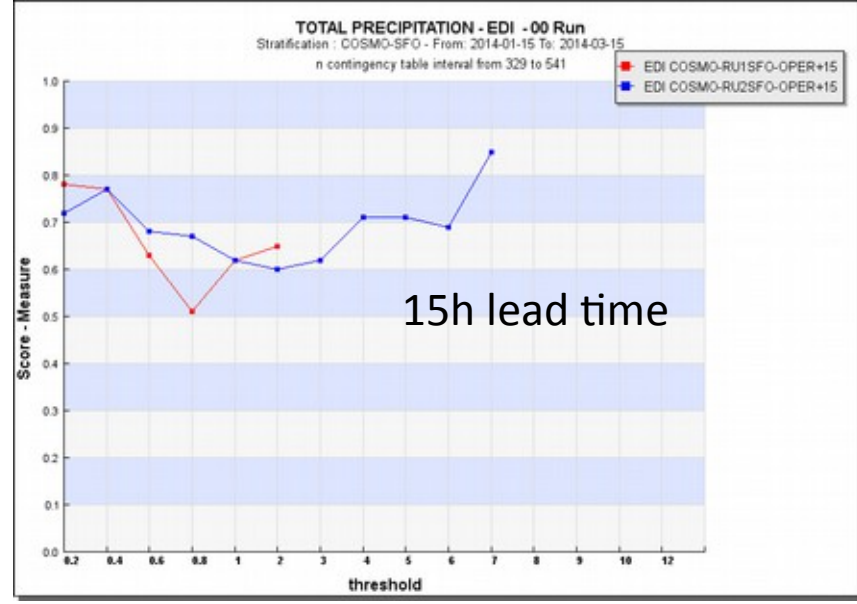
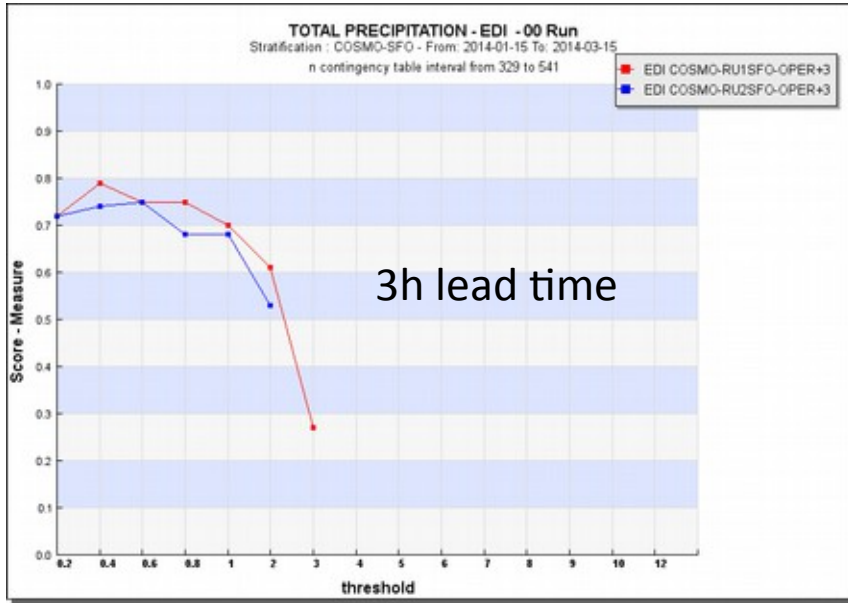


 COSMO-Ru2 domain (2.2-km resolution)

 COSMO-Ru1 domain (1.1-km resolution)

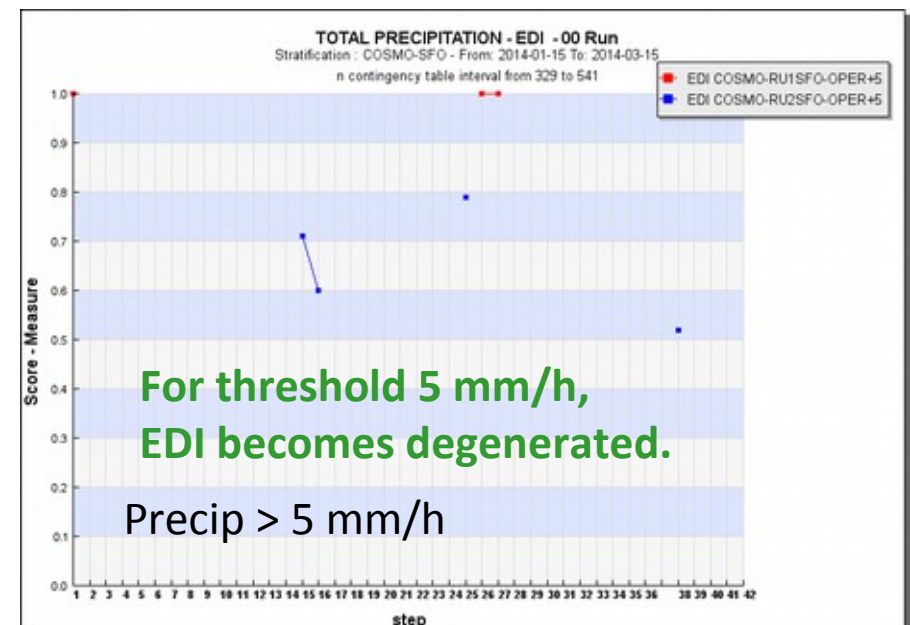
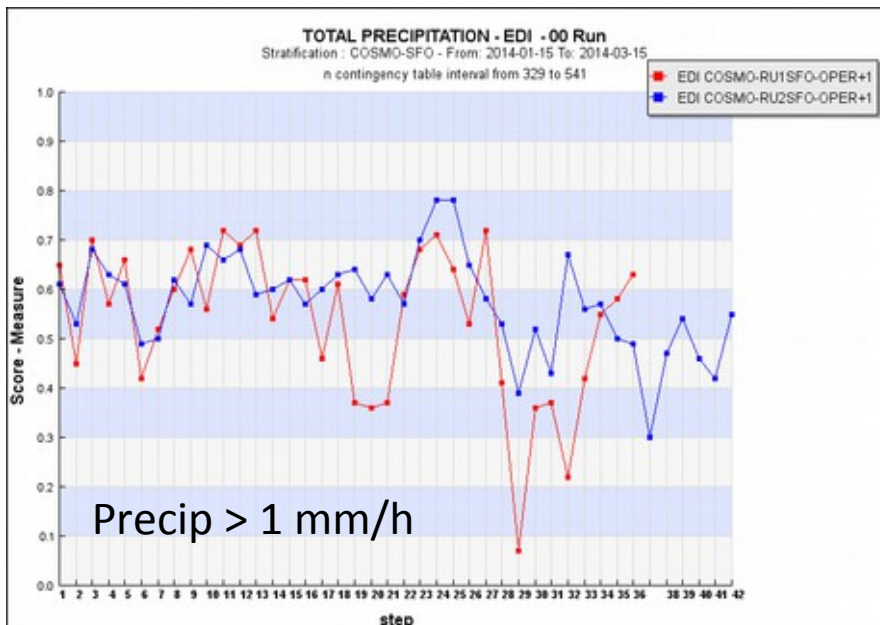
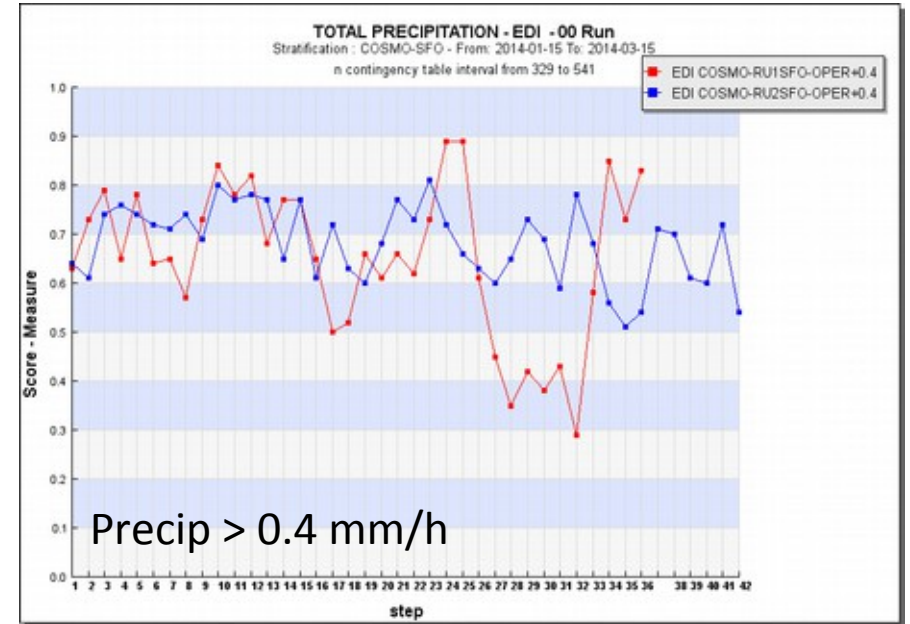
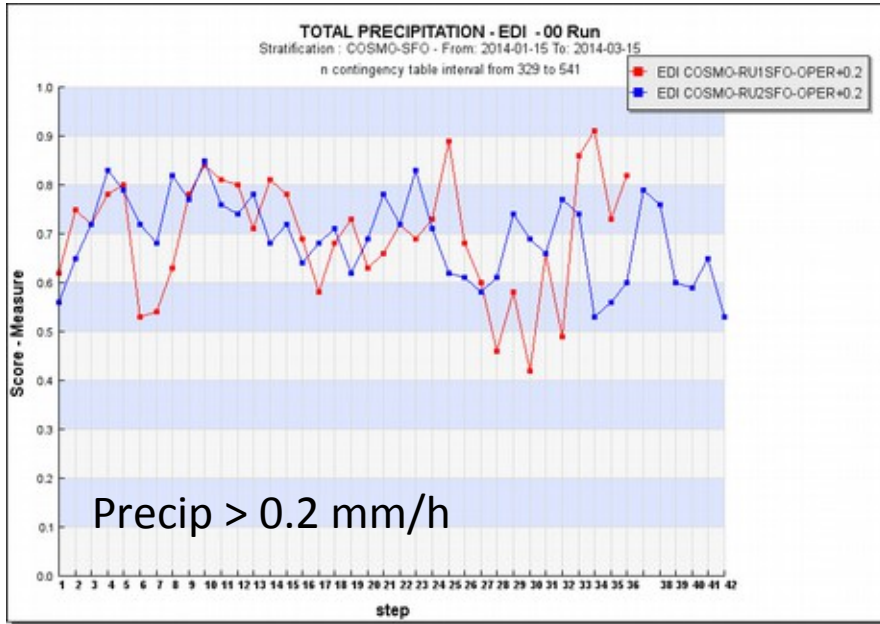


# EDI of 1h precipitation, Sochi region, Comparison with the station data (~23 stations) as a function of threshold, COSMO-Ru1 and COSMO-Ru2





# EDI of 1h precipitation, Sochi region, Comparison with the station data (~23 stations) As a function of lead time, **COSMO-Ru1** and **COSMO-Ru2**



Experiments on matching  
precipitation objects,  
Sochi region

# Setup of experiments

- Motivation: To apply CRA (Contiguous Rain Area, Ebert and McBride 2000) method (one of the object-based spatial verification methods) in order to avoid double-penalty problem of point-wise verification
- 1h precipitation fields from COSMO-Ru2 and COSMO-Ru1 compared to Sochi radar data
- 35 cases with intense precipitation over the period from 15 Jan to 15 Mar 2014 were considered
- Precipitation threshold for object identification: 1 mm/h, convolution smoother
- R SpatialVx package (developed by Eric Gilleland) for object identification and matching
- Problem: radar data cut-off areas

# Object-based methods

- In each approach, it is necessary to determine the distance between two objects, which is not always straightforward because of widely varying shapes and sizes of objects
- Once objects have been identified, it is often necessary to subsequently merge some objects together (that are part of the same weather system) (Gilleland et al. 2008)

# Functions for

## matching objects in R SpatialVx

- **Minboundmatch** (in single matches mode): each object is paired to only one object according to the smallest minimum boundary separation
- **Deltamm** – merges and/or matches using Baddeley's Delta Image Metric (Gilleland 2008), which yields a type of average pixel distance between sets (objects)  $A$  and  $B$  under certain assumptions (chosen in this case)
- **Centmatch** is similar to *deltamm*. It is based on the method proposed by Davis et al. (2006a). It is possible for more than one object to be matched to the same object in another field.

Objects are matched, if the centroid distance  $D$  is less than

- 1) the sum of the sizes of the two objects in question (size is the square root of the area of the object) ( $D=1$ )
- 2) the average size of the two objects in question ( $D=2$ )
- 3) a given constant ( $D=3$ )

**$D=1$  and  $D=2$  were used.**

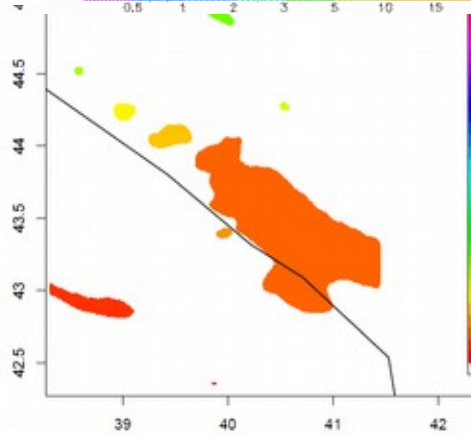
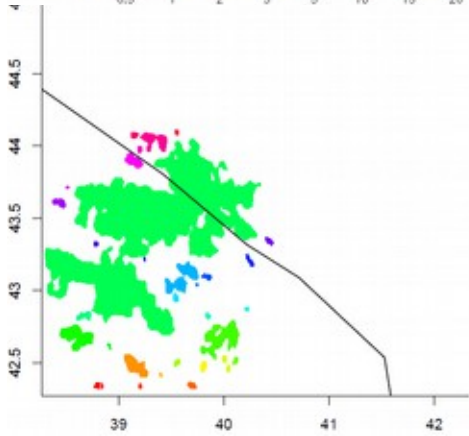
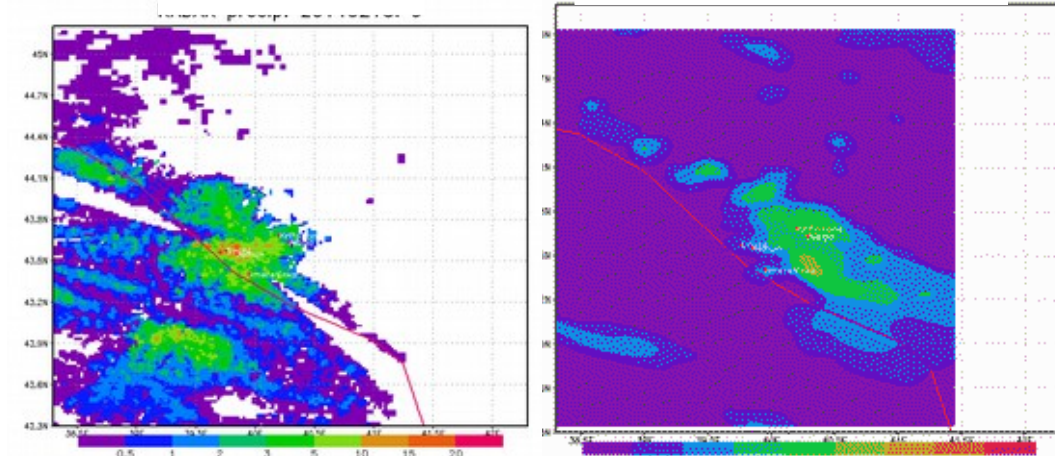
*Centmatch* doesn't merge objects explicitly, but determines possible merges applied if **MergeForce** function is run after *centmatch* (**used in this case**)



# Example: 2014021809, Identification of objects

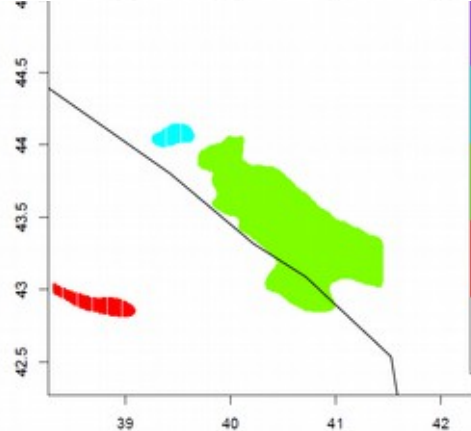
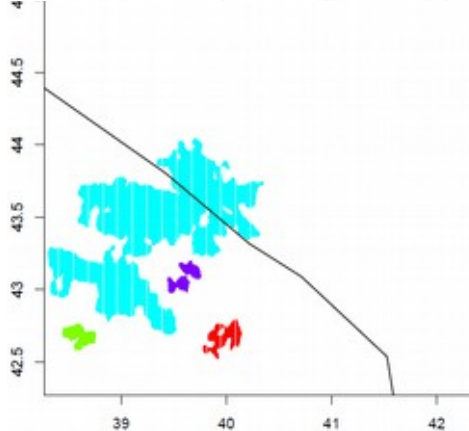
RADAR

COSMO-Ru2



New R SpatialVx option allowing cutting off too small objects proved very useful

NO cutting off



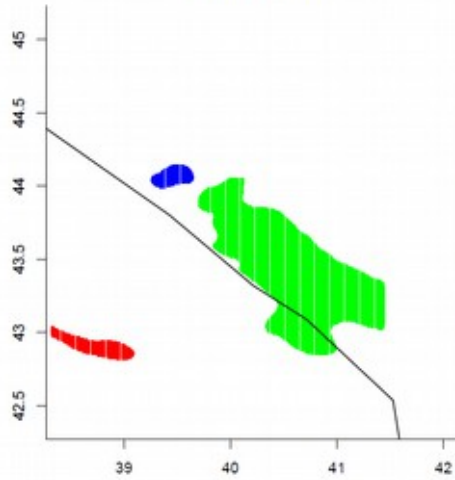
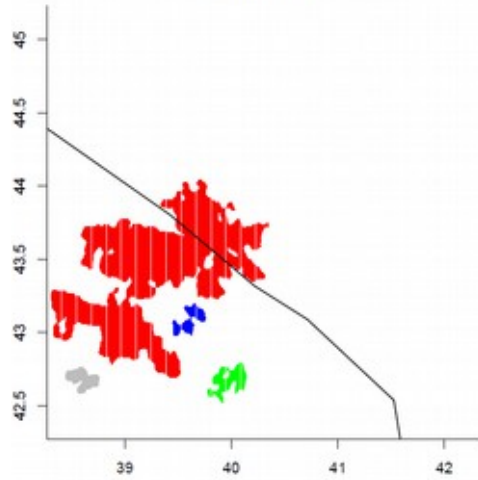
WITH cutting off

# Example: Object merging and matching

Minboundmatch

Akhun-RADAR  
Feature Field

Sochi  
Feature Field

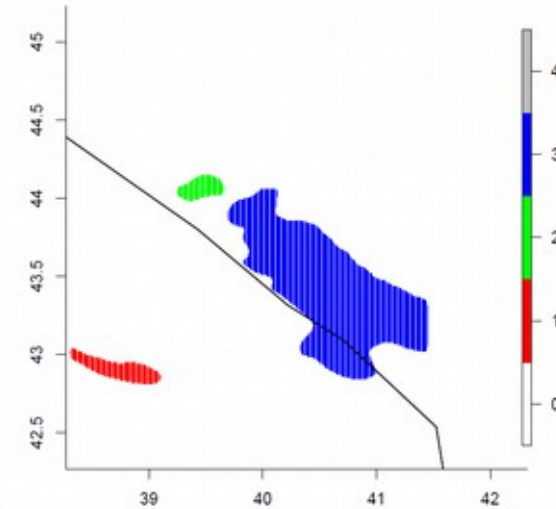
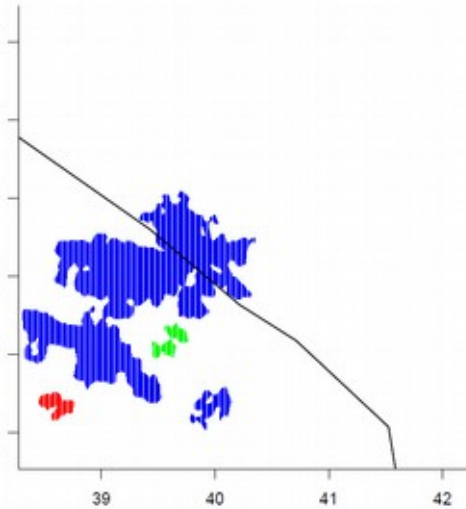


Centmatch, D=1

Akhun-RADAR  
Feature Field

Deltamm

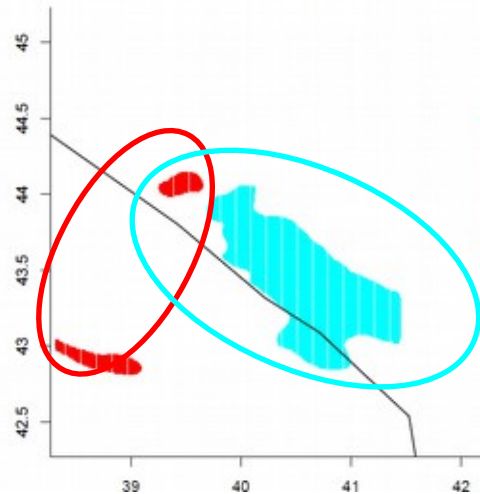
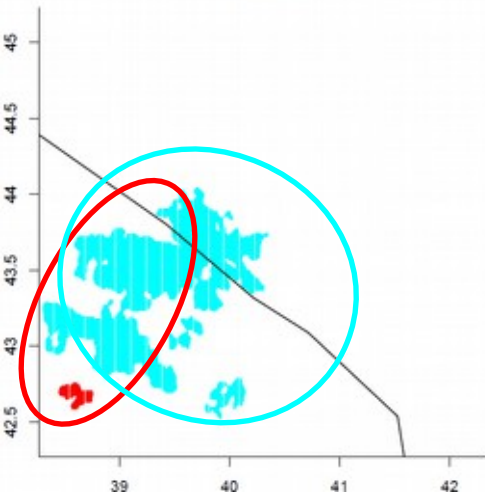
Sochi  
Feature Field



Centmatch, D=2

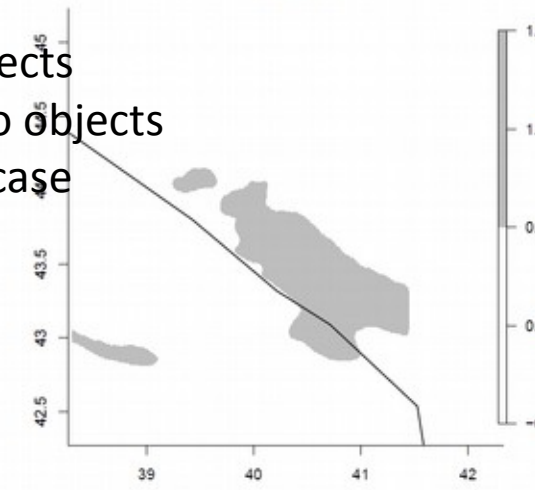
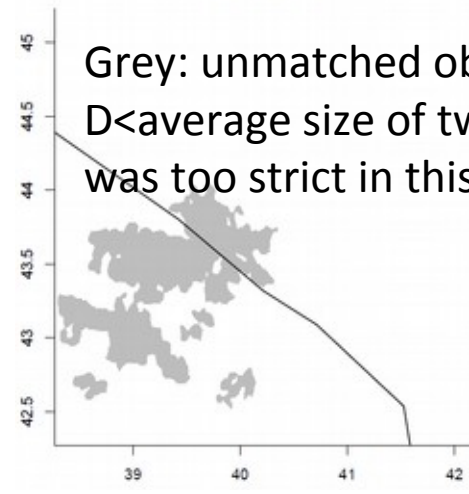
Akhun-RADAR  
Feature Field

Sochi  
Feature Field



Akhun-RADAR  
Feature Field

Sochi  
Feature Field



Grey: unmatched objects  
D < average size of two objects  
was too strict in this case

# Discussion

- It is difficult as yet to choose a best universal object matching method in the complex terrain
- Minboundmatch tends to find the first good match, the others can be less obvious. Sometimes, big objects are matched to small ones
- Matching based on centroid distance less than the sum of the sizes of two objects provides too much matches overall,
- Matching based on centroid distance less than the average size of two objects gives better results, however, leaving many unmatched objects