



# Supplement of

# Radar-based characterisation of heavy precipitation in the eastern Mediterranean and its representation in a convection-permitting model

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## Supplementary material

#### S1. Fractions Skill Score (FSS) statistic

The fractions skill score (FSS; Roberts and Lean, 2008) statistic is defined for each rainfall threshold (q) using a binary field (I) that equals 1 wherever pixel values are  $\geq q$ , and 0 elsewhere. Thus, the fraction of radar-derived (observed) pixels for a given rainfall threshold over a given neighbourhood length n (i.e., spatial averaging), termed  $O_n$ , and the similar modelled fraction derived from the Weather Research and Forecasting (WRF) model  $(M_n)$ , are used to calculate the mean square error (MSE) as follows:

(Eq. S1) 
$$MSE_n \equiv \overline{(O_n - M_n)^2}$$
,

where the overbar denotes averaging. The MSE is then used to calculate the FSS:

(Eq. S2) 
$$FSS_n \equiv \frac{MSE_n - MSE_{(n)ref}}{MSE_{(n)perfect} - MSE_{(n)ref}} = 1 - \frac{MSE_n}{MSE_{(n)ref}}$$

where  $MSE_{(n)perfect} \equiv 0$  is the MSE of a perfect forecast, and  $MSE_{(n)ref} \equiv \overline{O_n^2} + \overline{M_n^2}$ . The uniform FSS is defined as halfway between a random forecast and a perfect skill forecast:

(Eq. S3) 
$$FSS_{(n)uniform} \equiv \frac{1+f_{(n)0}}{2}$$
,

where  $f_0$  is the observed frequency, i.e., the fraction of observed pixels exceeding the threshold over the entire domain using a neighbourhood length of size n.

#### S2. Structure–amplitude–location (SAL) analysis

The structure-amplitude-location analysis (SAL; Wernli et al., 2008) shown in the text also requires setting up a rainfall threshold (f) that enables distinguishing precipitation objects that are greater than this threshold. Following is a summary of the calculation of each of the three components of SAL.

A component (amplitude):

(Eq. S4) 
$$A = \frac{\overline{R_M} - \overline{R_O}}{\frac{1}{2}(\overline{R_M} + \overline{R_O})}$$

where *R* is the rainfall accumulation field and *M* and *O* denote modelled (WRF) and observed (radar) rain, respectively, and  $A \in [-2,2]$ .

The *L* component (location) is the sum of two components. The first one  $(L_1)$  is a normalised measure of the distance between the centre of mass of the modelled and observed rain fields, and the second  $(L_2)$ 

considers the average distance between the centre of mass of the total precipitation fields and individual precipitation objects within them, as follows:

(Eq. S5) 
$$L_1 = \frac{|x_M - x_O|}{d}$$
,

where x denotes the centre of mass of a rain field and d is the largest possible geographical distance along the considered domain.

The second location component  $(L_2)$  weights each precipitation object using its total amount of rain  $(R_n)$ and a weighted average distance (r):

(Eq. S6) 
$$r = \frac{\sum_{n=1}^{M} R_n |x - x_n|}{\sum_{n=1}^{M} R_n}$$
,

where *n* is an index of precipitation objects ranging from 1 to the number of objects existing (*M*).  $L_2$  is computed through the difference between the modelled distance ( $r_M$ ) and the observed one ( $r_O$ ), calculated according to Eq. S7, for the modelled and observed precipitation objects, respectively.

(Eq. S7) 
$$L_2 = 2(\frac{|r_M - r_O|}{d})$$

Finally, the *L* component is simply the sum of  $L_1$  and  $L_2$ :

(Eq. S8) 
$$L = L_1 + L_2$$

where  $L \in [0,2]$ .

The S component (structure) is calculated through a scaled volume of each precipitation object  $(V_n)$ :

(Eq. S9) 
$$V_n = \frac{R_n}{R_n^{max}}$$
,

where  $R_n^{max}$  is the maximum rainfall value of the precipitation object n. The weighted mean of the scaled volume is calculated through:

(Eq. S10) 
$$V = \frac{\sum_{n=1}^{M} R_n V_n}{\sum_{n=1}^{M} R_n},$$

which is then used to calculate the S component:

(Eq. S11) 
$$S = \frac{V_M - V_O}{\frac{1}{2}(V_M + V_O)},$$

where  $V_M$  and  $V_O$  represent the scaled volume calculated using the modelled and observed rain fields, respectively, and  $S \in [-2,2]$ .

# S3. Heavy precipitation events (HPEs) identified and analysed

	Start time*	End time*	Synoptic	HPE duration [h]						
П <b>Г</b> С #			classification <sup>#</sup>	1	3	6	12	24	48	72
1	2-11-1991 9:00	5-11-1991 9:00	MC	Х	Х	Х	Х	Х	Х	Х
2	22-2-1992 8:00	27-2-1992 21:00	MC						Х	Х
3	23-11-1992 9:00	26-11-1992 7:00	MC	Х	Х	Х	Х			
4	12-12-1992 14:00	18-12-1992 13:00	MC	Х	Х	Х	Х	Х	Х	Х
5	31-3-1993 9:00	2-4-1993 2:00	ARST			Х				
6	21-12-1993 12:00	23-12-1993 15:00	ARST						Х	
7	21-2-1994 19:00	25-2-1994 0:00	MC		Х	Х	Х			
8	1-11-1994 15:00	7-11-1994 13:00	ARST	Х	Х	Х	Х	Х	Х	Х
9	14-11-1994 1:00	18-11-1994 5:00	MC	Х	Х	Х	Х	Х	Х	Х
10	15-12-1994 12:00	20-12-1994 21:00	MC	Х	Х			Х	Х	Х
11	28-12-1994 10:00	31-12-1994 23:00	MC			Х				
12	4-2-1995 8:00	9-2-1995 10:00	MC					Х	Х	Х
13	1-11-1995 11:00	3-11-1995 14:00	MC	Х	Х					
14	7-11-1995 10:00	10-11-1995 17:00	MC							Х
15	6-3-1996 13:00	8-3-1996 4:00	MC		Х	Х	Х	Х	Х	
16	11-12-1996 14:00	14-12-1996 15:00	ARST	Х	Х	Х	Х	Х	Х	Х
17	13-1-1997 11:00	17-1-1997 7:00	MC	Х	Х	Х	Х	Х	Х	Х
18	3-3-1997 6:00	4-3-1997 16:00	MC	Х	Х	Х	Х			
19	19-10-1997 11:00	20-10-1997 10:00	MC		Х	Х	Х			
20	25-11-1997 10:00	27-11-1997 9:00	ARST		Х	Х	Х	Х		
21	4-4-1998 4:00	4-4-1998 17:00	MC			Х	Х			
22	28-12-1998 6:00	31-12-1998 21:00	MC	Х	Х	Х	Х	Х		
23	13-12-1999 6:00	15-12-1999 8:00	MC	Х	Х	Х	Х			
24	18-1-2000 6:00	24-1-2000 2:00	MC		Х	Х	Х	Х	Х	Х
25	25-1-2000 15:00	28-1-2000 20:00	MC	Х	Х	Х	Х	Х	Х	Х
26	12-2-2000 22:00	16-2-2000 16:00	MC	Х	Х	Х	Х	Х	Х	Х
27	29-11-2000 0:00	1-12-2000 10:00	MC				Х	Х	Х	Х
28	19-12-2000 6:00	21-12-2000 17:00	MC	Х	Х					
29	30-4-2001 9:00	2-5-2001 17:00	MC	Х	Х	Х	Х	Х	Х	Х
30	9-12-2002 6:00	12-12-2002 6:00	MC		Х	Х	Х	Х	Х	Х
31	2-1-2003 16:00	4-1-2003 12:00	MC		Х	Х				
32	27-1-2003 10:00	30-1-2003 13:00	MC	Х	Х	Х	Х	Х		Х
33	3-2-2003 0:00	5-2-2003 16:00	MC				Х	Х	Х	
34	17-2-2003 19:00	22-2-2003 23:00	MC						Х	Х
35	24-2-2003 1:00	28-2-2003 2:00	MC				Х	Х	X	X

Table S1 – HPEs identified and analysed in this study

36	1-12-2003 14:00	5-12-2003 20:00	ARST	Х	Х	Х	Х	Х	Х	Х
37	14-12-2003 16:00	15-12-2003 10:00	MC	Х	Х	Х	Х	Х		
38	15-12-2005 15:00	18-12-2005 9:00	MC		Х	Х	Х			
39	18-12-2007 14:00	21-12-2007 8:00	MC			Х	Х	Х	Х	
40	2-1-2008 3:00	5-1-2008 19:00	MC		Х	Х	Х	Х	Х	Х
41	17-1-2010 16:00	22-1-2010 6:00	MC	Х	Х	Х	Х	Х	Х	Х

\*Local winter time (UTC+2), presented as day-month-year and hour.

\*Simplified synoptic classification (Sect. 3.4 in the main text).

### S4. WRF namelist.input file example

&time_control	
run_days	= 0,
run_hours	= 138,
run_minutes	= 0,
run_seconds	= 0,
start_year	= 2010, 2010, 2010,
start_month	= 01, 01, 01,
start_day	= 16, 16, 16,
start_hour	= 12, 12, 12,
start_minute	= 00, 00, 00,
start_second	= 00, 00, 00,
end_year	= 2010, 2010, 2010,
end_month	= 01, 01, 01,
end_day	= 22, 22, 22,
end_hour	= 06, 06, 06,
end_minute	= 00, 00, 00,
end_second	= 00, 00, 00,
interval_seconds	= 21600
input_from_file	= .true.,.true.,.true.,
history_interval	= 360, 180, 10,
frames_per_outfile	= 1, 1, 1,
restart	= .false.,
restart_interval	= 1440,
io_form_history	= 2,

io\_form\_restart = 2,

io\_form\_input = 2,

io\_form\_boundary = 2,

= 0,

debug\_level

iofields\_filename = "varsNot2Use\_d01.txt", "varsNot2Use\_d02.txt", "varsNot2Use\_d03.txt",

ignore\_iofields\_warning = .true.,

/

&domains

time_step	= 8,
time_step_fract_nur	n = 0,
time_step_fract_den	= 1,
use_adaptive_time_s	tep = .true.,
step_to_output_time	e = .true.,
target_cfl	= 1.2, 1.2,1.2,
target_hcfl	= .84, .84,0.84,
max_step_increase_	pct = 5, 51,51,
starting_time_step	= -1, -1,-1,
max_time_step	= -1, -1,-1,
min_time_step	= -1, -1,-1,
adaptation_domain	= 1,
max_dom	= 3,
e_we = 100, 2	221, 551,
e_sn = 100, 2	221, 551,
e_vert	= 68, 68, 68,
p_top_requested	= 2500,
num_metgrid_levels	= 61,
num_metgrid_soil_le	evels = 4,
dx	= 25000, 5000, 1000,
dy	= 25000, 5000, 1000,
grid_id	= 1, 2, 3,
parent_id	= 0, 1, 2,
i_parent_start = 1	, 28, 55,

j\_parent\_start = 1, 28, 55, parent\_grid\_ratio = 1, 5, 5, parent\_time\_step\_ratio = 1, 5, 5, feedback = 1, smooth\_option = 0, /

### &physics

mp_physics				= 8, 8, 8,
cu_physics				= 6, 6, 0,
ra_lw_physics				= 4, 4, 4,
ra_sw_physics				= 4, 4, 4,
bl_pbl_physics				= 2, 2, 2,
sf_sfclay_physics				= 2, 2, 2,
sf_surface_physics				= 2, 2, 2,
radt	= 15,	15,	1	5,
bldt	= 0,	0,	0,	
cudt	= 2,	2,	2,	
icloud	= 1,			
isfflx				= 1,
ifsnow				= 1,
num_soil_layers		= 4,		
num_land_cat		= 21,	,	
sf_urban_physics		= 0,	0	, 0,
surface_input_source	e			= 1,
/				

#### &fdda

/

#### &dynamics

w\_damping = 1,

diff_opt	= 1, 1, 1,
km_opt	= 4, 4, 4,
diff_6th_opt	= 0, 0, 0,
diff_6th_factor	= 0.12, 0.12, 0.12,
base_temp	= 290.
damp_opt	= 3,
zdamp	= 5000., 5000., 5000.,
dampcoef	= 0.2, 0.2, 0.2
khdif	= 0, 0, 0,
kvdif	= 0, 0, 0,
epssm	= 0.2, 0.2, 0.2,
non_hydrostatic	= .true., .true., .true.,
moist_adv_opt	= 1, 1, 1,
scalar_adv_opt	= 1, 1, 1,
gwd_opt	= 1,
/	

# &bdy\_control

spec_bdy_width	= 5,
spec_zone	= 1,
relax_zone	= 4,
specified	= .true., .false.,.false.,
nested	= .false., .true., .true.,
/	

# &grib2

# /

&namelist\_quilt nio\_tasks\_per\_group = 0, nio\_groups = 1, /

### S5. Figures



Figure S1: An example of the spatial autocorrelation analysis (Sects. 3.5.4, 4.4.2 in the main text). The left panel shows a 10-min rainfall map based on radar data from HPE #1. The right panel shows the 2-D autocorrelation field of the same map. The red ellipse represents the approximate  $e^{-1}$  correlation region and its axes are in black. Deviation of the major axes from the east-west axis (grey) is denoted  $\alpha$ . The short-to-long axis ratio defines the ellipticity of the autocorrelation field.



Figure S2: Accumulated precipitation (convective [RAINC] + non-convective [RAINNC] rainfall) in the coarsest WRF domain during HPE #5 (Table S1) and the approximate range of the Shacham radar (Fig. 1 in the main text). Notice the absence of rainfall within the radar range, as opposed to the radar QPE (Fig. 8a in the main text).