

1 **Sea breeze thunderstorms in the eastern Iberian**
2 **Peninsula. Neighborhood verification of**
3 **HIRLAM and HARMONIE precipitation**
4 **forecasts**

5
6 Cesar Azorin-Molina^{1,*}, Sander Tijm², Elizabeth E. Ebert³, Sergio M. Vicente-Serrano¹,
7 María J. Estrela⁴

8 *Surnames (or family names) are underlined*

9
10
11 1. *Instituto Pirenaico de Ecología, Consejo Superior de Investigaciones Científicas (IPE-CSIC),*
12 *Departamento de Procesos Geoambientales y Cambio Global, Avda. Montañana 1005,*
13 *50059-Zaragoza, Spain*

14 2. *Royal Netherlands Meteorological Institute (KNMI), PO Box 201, NL-3730 AE De Bilt, The*
15 *Netherlands*

16
17 3. *Centre for Australian Weather and Climate Research (CAWCR), Bureau of Meteorology,*
18 *GPO Box 1289, VIC 3001 Melbourne, Australia*

19
20 4. *Laboratory of Meteorology-Climatology, Mixed Unity CEAM-UVEG, Department of*
21 *Physical Geography, University of Valencia, Avda. Blasco Ibáñez 28, 46010-Valencia, Spain*
22
23

24 * Corresponding author at: Instituto Pirenaico de Ecología, Consejo Superior de
25 Investigaciones Científicas (IPE-CSIC), Departamento de Procesos Geoambientales y
26 Cambio Global, Avda. Montañana 1005, 50059-Zaragoza, Spain. Tel.: (+34)
27 97.671.60.34 ; fax: (+34) 97.436.32.22

28 *E-mail address: cazorin@ipe.csic.es (C. Azorin-Molina)*

29
30
31 Manuscript published in Atmospheric Research
32
33

34 ABSTRACT

35

36 In this study we investigated sea breeze thunderstorms with intense convective activity
37 (i.e., heavy rainfall, hail and gusty winds) that occurred over the eastern Iberian
38 Peninsula (Spain) and were missed by the operational HIRLAM model. We used two
39 grid-spacing setups (5.0-km and 2.5-km) of the hydrostatic HIRLAM model, and the
40 non-hydrostatic spectral HARMONIE suite (2.5-km), to simulate isolated convection
41 associated with sea breezes. The overall aim is to estimate the ability of these three
42 experimental setups, in particular the HARMONIE model as the forthcoming
43 operational Numerical Weather Prediction in most European Weather Services, to
44 correctly simulate convective precipitation associated with sea breezes. We evaluated
45 high-resolution gridded precipitation forecasts from HIRLAM and HARMONIE suites
46 for 15 sea breeze thunderstorms against high-density gridded raingauge measurements
47 applying different neighborhood verification techniques. The results indicate that
48 higher horizontal resolutions of HIRLAM and HARMONIE models succeeded in
49 predicting the occurrence of these missed sea breeze thunderstorms, the HARMONIE
50 suite being the most capable of providing good estimates of accumulated precipitation
51 in convective events in terms of space and time. Advances in quantitative precipitation
52 forecasting of locally driven convection could have practical applications for
53 nowcasting dangerous sea breeze convective phenomena.

54 *Keywords:* sea breeze thunderstorm; operational forecasting; HIRLAM; HARMONIE;
55 neighborhood verification; Iberian Peninsula.

56

57 **1. Introduction**

58

59 The effect of sea breezes in triggering deep convection has been noted in
60 numerous coastal areas around the world (Simpson, 1994). Many numerical modelling
61 and observational studies, particularly for the subtropical Florida Panhandle (e.g.,
62 Pielke, 1974; Pielke and Cotton, 1977; Pielke and Mahrer, 1978; Blanchard and Lopez,
63 1985; Nichols et al., 1991; Wakimoto and Atkins, 1994), found that low-level sea
64 breeze convergence and consequently sea breeze front (SBF) development enhance
65 planetary boundary layer (PBL) air parcels to lift up to the level of free convection
66 (LFC) (Wilson, 2008; Muppa et al., 2012). Convective initiation can also occur from
67 horizontal convective rolls (HCR) ahead of the advancing SBF (Atkins et al., 1995).
68 Both SBF and HCR updrafts merge, promoting deep convective sea breeze
69 thunderstorms that can sometimes be extraordinarily severe, causing significant rainfall,
70 hail and gusty winds along the frontal boundary (Dailey and Fovell, 1999; Fovell and
71 Dailey, 2001; Fovell, 2005).

72 Convection resulting from horizontal low-level convergence along SBFs is
73 intensified by several factors (Simpson et al., 2007). The most relevant are as follows:
74 (a) anabatic valley wind circulations on heated south-facing mountain slopes play a
75 major role in strengthening uplift processes and developing deep convection by
76 combining with moist sea breezes that transport water vapor to the coastal mountains in
77 the daytime (Millan et al., 2005); (b) convergence and SBF intensification is also
78 enhanced by frictional effects (upslope orographic lifting) produced in coastal areas of
79 complex terrain (Petterssen, 1956; Pérez-Landa et al., 2006; Papanastasiou et al., 2010);
80 (c) strong low-level sea breeze convergence also occurs on convex coastlines (e.g.,

81 peninsulas, capes, points, and also islands) contributing to increased upward vertical
82 motion, *Cumulus* (Cu) and *Cumulonimbus* (Cb) activity and thunderstorms along the
83 frontal area (Neumann, 1951; Pielke, 1974; Purdom, 1976; Strickler, 2003); (d) a
84 sharply defined discontinuity and convergence intensification is observed inland along
85 the SBF under offshore large-scale synoptic flows (Bechtold et al., 1991; Atkins and
86 Wakimoto, 1997), and light to moderate winds aloft ($<5.0 \text{ m s}^{-1}$) result in more clouds at
87 the leading edge of sea breezes (Azorin-Molina et al., 2009); and (f) regions of high soil
88 moisture expect heavy precipitation along the SBF (Baker et al., 2001). Most of the
89 aforementioned factors interact over the complex eastern façade of the Iberian Peninsula
90 (Spain), where thermally induced local circulations interact.

91 Short-term forecasting of the timing, location and intensity of isolated sea breeze
92 thunderstorms represents a challenging task in numerical weather prediction (NWP),
93 mainly due to the uncertainties in the initial conditions, limited knowledge about the
94 cloud microphysical processes, and difficulties in resolving low-level sea breeze
95 convergence and convection with fairly coarse horizontal resolution operational NWP
96 models (Mazarakis et al., 2009). Furthermore, isolated sea breeze thunderstorm cells
97 with severe weather can develop unexpectedly under weakly defined synoptic-scale or
98 mesoscale precursor disturbances (Wilson, 2008), and may be missed by forecasts.
99 Strong low-level sea breeze boundaries can deliver enough energy to overcome the
100 stable cap of the PBL and generate local showers and thunderstorms unexpectedly. For
101 instance, deep convection associated with sea breezes can occur even when sounding
102 indices indicate stable weather conditions. This is because sounding indices do not
103 consider layers below 850 hPa, where strong low level convergence may accumulate
104 lower tropospheric moisture (Pielke et al., 1971) or capping of inversions can occur,
105 enhancing or weakening lifting mechanisms, respectively.

106 Low-level sea breeze convergence occurs preferentially during the warm season
107 (May-September), but also in the transition months of April and October, over the
108 complex eastern façade of the IP. Its time of occurrence is nearly always in the mid-
109 afternoon. Despite the high level of occurrence of thermally-driven winds (sea breezes
110 blow two out of three days of the year, Azorin-Molina and Martin-Vide, 2007) and also
111 that sea breeze convection brings an average of 100-125 mm yearly to inland areas
112 during the summer dry season (Millan et al., 2005), there is little knowledge concerning
113 the important role that sea breezes play in convection initiation in the eastern coast of
114 Spain. A review of these few investigations was presented by Azorin-Molina et al.
115 (2009), who used high-resolution cloud frequency composites derived from NOAA-
116 AVHRR data to identify the location of five preferential sea breeze convergence zones
117 (SBCZ; hot spots) in the Iberian Mediterranean area and the Isle of Mallorca. The
118 current study is focused on two of them, i.e., the SBCZ2 (eastern region of the Iberian
119 system mountains; 1000-1900 m) and the SBCZ3 (Prebetic mountain ranges; 1000-
120 1600 m). These regions correspond to the east of the Iberian Peninsula, an
121 orographically highly complex area (Figure 1a).

122 The main goal of this study is to estimate the ability of NWP to correctly
123 simulate convective precipitation associated with sea breezes. Two different setups
124 (5.0-km and 2.5-km horizontal grid-spacing) of the operational HIRLAM model and
125 the non-hydrostatic spectral HARMONIE suite (2.5-km horizontal grid-spacing; Hirlam
126 Aladin Regional/Meso-scale Operational NWP In Europe) are evaluated. Three
127 different neighborhood (also known as ‘fuzzy’) verification techniques are applied here
128 in order to measure the strength of the HIRLAM and HARMONIE agreement with the
129 observations. The article is structured as follows. In section 2, we summarize the model
130 description, set up and initialization of HIRLAM and HARMONIE suites, present the

131 sea breeze thunderstorms simulated and observed precipitation data, and briefly
132 describe the neighborhood verification methods applied in this study. In section 3, the
133 performance of HIRLAM and HARMONIE gridded precipitation forecasts is evaluated
134 against gridded precipitation observations. In section 4, the sea breeze thunderstorm
135 occurred on 7 August 2008 is analyzed. Finally, a summary and discussion of the
136 findings from this study are presented in section 5.

137

138 **2. Data and methods**

139 *2.1. HIRLAM and HARMONIE model description, set up and initialization*

140 The NWP systems used here are the three-dimensional hydrostatic grid-point
141 model version 7.2.2 of HIRLAM (Undén et al., 2002), and the non-hydrostatic spectral
142 model version 36h1.2 of HARMONIE (Seity et al., 2011). The HIRLAM short-range
143 forecasting model was chosen for this research because it is currently employed as one
144 of the most important operational NWP system at the AEMET, and by eight other
145 European Weather Services: Denmark, Estonia, Finland, Iceland, Ireland, the
146 Netherlands, Norway and Sweden. The HARMONIE limited area model is also used in
147 this study because it will be implemented shortly as the operational NWP in most of the
148 European Weather Services. Actually, it is currently being evaluated as a next-
149 generation replacement for HIRLAM (van der Plas et al., 2012; de Bruijn and de Rooy,
150 2012).

151 The dynamical core of HIRLAM model is based on a semi-implicit semi-
152 Lagrangian discretization of the multi-level primitive equations, employing a hybrid
153 coordinate in the vertical. The comprehensive set of physical parameterization schemes

154 selected in the HIRLAM model suite in order to take into account a variety of sub-
155 gridscale physical processes include: (a) a radiation scheme (Savijärvi, 1990), (b) an
156 adapted Rasch-Kristjansson condensation (Zhang et al., 2003; Ivarsson, 2007) and a
157 Kain-Fritsch mass-flux convection scheme with CAPE closure (Kain, 2004; Calvo,
158 2007) , (c) a prognostic moist turbulent kinetic energy (TKE) parameterization (Tijm
159 and Lenderink, 2003), (d) a tiled surface approach distinguishing seven surface types
160 (Interaction Soil-Biosphere-Atmosphere, ISBA surface scheme, Noilhan and Planton,
161 1989; snow and forest, Gollvik, 2002, 2004; lake, Mironov, 2008, Kourzeneva et al.,
162 2008), and (e) a mean and subgrid-scale orography parametrization (Rontu et al., 2002).
163 The HIRLAM system also includes a 4D-VAR data assimilation system (Huang et al.,
164 2002) with analysis of conventional and non-conventional observations. More detailed
165 descriptions about the dynamical and numerical features of the HIRLAM short-range
166 forecasting model can be found in Undén et al. (2002) and at the following URL:
167 [http://hirlam.org/index.php?option=com_content&view=article&id=64:general-
168 description-of-the-hirlam-model&catid=48:synoptic-scale-model-hirlam&Itemid=101](http://hirlam.org/index.php?option=com_content&view=article&id=64:general-
168 description-of-the-hirlam-model&catid=48:synoptic-scale-model-hirlam&Itemid=101)
169 (last accessed 1 November 2013).

170 The HARMONIE model suite is designed in a cooperation between the
171 HIRLAM and ALADIN (*Aire Limitée Adaptation dynamique Développement*
172 *InterNational*) consortia. The dynamical core of the mesoscale model HARMONIE is
173 developed by ALADIN high precision short-range NWP model and based on a two-
174 time level semi-implicit semi-Lagrangian discretization of the fully elastic equations,
175 also employing a hybrid coordinate in the vertical. ALADIN is a spectral limited-area
176 model developed at Météo-France and based on the ARPEGE/IFS model. The
177 HARMONIE mesoscale option consists of a combination of ALADIN's non-hydrostatic
178 dynamics and Mesoscale Non-Hydrostatic atmospheric model (NH;

179 <http://mesonh.aero.obs-mip.fr/mesonh410>; last accessed 1 November 2013) physics.
180 The parametrizations of a variety of sub-gridscale physical processes are the same as
181 those adopted of the AROME physics (Seity et al., 2011). The HARMONIE system
182 includes a 3D-VAR data assimilation system. A more detailed description of the
183 HARMONIE model can be found at this URL:
184 http://hirlam.org/index.php?option=com_content&view=article&id=65&Itemid=102
185 (last accessed 1 November 2013).

186 Here, we are interested in evaluating the precipitation performance of the
187 HIRLAM 0.05° (4.3-km length; hereafter HIR-D5) horizontal grid-spacing against a
188 higher horizontal grid-resolution of HIRLAM 0.025° (2.5-km length; hereafter HIR-
189 D2.5) and HARMONIE 0.025° (2.5-km length; hereafter HAR-D2.5), for those sea
190 breeze thunderstorm episodes described in section 2.2. On the one hand, the HIRLAM
191 model is set up with unique domains with an integration area of 300 x 306 horizontal
192 grid points (latitude by longitude) and the size of 1495 km x 1525 km for the HIR-D5
193 experiment, and 400 x 406 horizontal grid points and 997.5 km x 1012.5 km for the
194 HIR-D2.5 experiment. Increased boundary layer vertical resolution can provide
195 improved forecast performance (Ries and Heinke-Schlunzen, 2009), and therefore the
196 vertical domain for the two set ups of the HIRLAM contains 60 sigma levels with
197 enhanced resolution (27 levels in the lowest 3 km) in the PBL. The model top is at 10
198 hPa and the lowest model level is around 32 m. On the other hand, the HAR-D2.5
199 experiment is set up with an integration area of 400 x 400 horizontal grid points
200 (latitude by longitude) and the size of 1000 km x 1000 km, also employing a vertical
201 domain of 60 hybrid levels with enhanced resolution (27 levels in the lowest 3 km) in
202 the PBL. The HARMONIE model top is at 1 hPa and the lowest model level is around
203 10 m. The ECMWF archive supplies the operational boundaries data for both HIRLAM

204 and HARMONIE suites with a horizontal grid-spacing of $0.5^\circ \times 0.5^\circ$ and a temporal
205 resolution of 6 h. These boundaries are also applied as initial conditions. Lateral
206 boundaries are overspecified, all variables being externally prescribed by the nesting
207 model. A relaxation zone of 10 grid points is normally adopted and boundary relaxation
208 is performed after the horizontal diffusion. Additionally, at the upper boundary a
209 condition of zero vertical velocity is imposed. The dynamics time step is 120 s for the
210 HIR-D5, and 60 s for the HIR-D2.5 and the HAR-D2.5 experiments. All the
211 simulations were initialized at 1200 UTC the day before the sea breeze thunderstorm
212 event, and integrated for 48 h until 1200 UTC the day after the episode, with model
213 outputs at intervals of 1 h. The domains for each NWP system are represented in Figure
214 1b.

215

216 *2.2. Simulated sea breeze thunderstorms and observed precipitation data*

217 We exclusively focused on 15 deep convective sea breeze thunderstorms that
218 were missed by the operational HIRLAM model; i.e., ‘observed but not forecasted
219 episodes’. A detailed verification of ‘forecasted but not observed events’ (i.e., false
220 alarms) is also an interesting topic for future work, but is out of the scope of this study
221 aimed in analyzing the ability of hydrostatic and non-hydrostatic convection-permitting
222 models in resolving these sea breeze thunderstorms.

223 These thunderstorms corresponded to typical sea breeze fronts that developed
224 explosively under weakly defined synoptic-mesoscale forcing; i.e., weak surface
225 pressure gradient in the western Mediterranean basin. Some of these storms were
226 severe, causing significant amounts of rainfall, hail and gusty winds. This synoptic

227 pattern allowed the inflow of moisture driven by well-developed local thermal
228 circulations, which mainly triggered these storms at the low-level convergence zone of
229 sea breezes. All these convective events were isolated and stationary as was revealed in
230 the geostationary and polar satellite images, Doppler radar, and lightning data. Table I
231 summarizes the list of sea breeze thunderstorms simulated, the number of raingauge
232 station points used for the neighborhood verification, and the maximum of precipitation
233 recorded for each episode at 06 UTC.

234 With the aim to evaluate the performance of HIRLAM *versus* HARMONIE
235 precipitation forecasts of these sea breeze thunderstorm episodes, this research used 24
236 h-precipitation accumulations (from 0800 UTC till 0800 UTC) from a high-density
237 network of Hellman-type raingauges covering the eastern region of the IP (Figure 1b).
238 The observed precipitation data have been supplied mainly by the AEMET, but we also
239 used a small number of observations from automatic weather stations (AWS; tipping
240 bucket raingauges) collected from the Mediterranean Center for Environmental Studies
241 (CEAM; http://www.ceam.es/index_i.htm; last accessed 1 November 2013), the
242 Automatic System of Hydrological Information of the Ebro basin management agency
243 (SAIH; <http://www.saihebro.com/saihebro/index.php>; last accessed 1 November 2013),
244 the Valencian Institute for Agriculture Research (IVIA; <http://www.ivia.es/>; last
245 accessed 1 November 2013), the Agro-climatic Information System for Irrigation
246 (SIAR, <http://crea.uclm.es/siar/>; SIAM, <http://siam.imida.es/>; last accessed 1 November
247 2013) and the METEOCLIMATIC network (<http://www.meteoclimatic.com/>; last
248 accessed 1 November 2013). The high-density raingauge network consists of a
249 minimum of 304 and a maximum of 533 observation points depending on each episode
250 and covers an area bounded between 37°00'N and 41°00'N and 3°00'W and 1°00'E, i.e.
251 both sea breeze convergence zones (SBCZ2 and SBCZ3; see Fig. 1a) with intensified

252 convective activity. Since data were obtained from various institutions, the 24-h raw
253 accumulated precipitation passed quality control (Aguilar et al., 2003) based on
254 removing gross errors due to archiving, transcription and digitalization. This basically
255 consisted on screening out suspicious precipitation values (outliers), such as extreme
256 and large values for 10-minute, 30-minute or daily measurements.

257 The precipitation forecasts and observations were gridded onto a regular grid
258 with dimensions of 81x81 grid points for the 5-km runs, and 161x161 for the 2.5-km
259 runs, resulting in a grid forecast and grid observation with the same dimensions for each
260 episode and model run. For gridding the observations into a regular grid we placed the
261 point observation values at the grid points closest to the observation points. When more
262 than one observation was present in a grid box of the single intercomparison grid, which
263 occurred for only two grid points in the total 15 sea breeze thunderstorms, the highest
264 value was chosen. Here we compared 24 h-precipitation accumulations (from 08:00
265 UTC till 08:00 UTC) from the gridded simulated and observed precipitation boxes.

266

267 *2.3. Neighborhood or 'fuzzy' verification methods*

268 The statistical verification of high-resolution gridded precipitation forecasts such
269 as those from HIRLAM and HARMONIE model suites against raingauge data is
270 increasingly conducted by applying neighborhood or 'fuzzy' methods. We decided to
271 use this approach because traditional verification metrics (e.g., mean square error, root
272 mean square error, etc.) typically give poor and uninformative values for high resolution
273 forecasts of small-scale features (e.g., low-level convergence of sea breezes triggering
274 isolated convection) since exact matches between forecasts and observations are quite

275 difficult to achieve (Ebert, 2008). However, neighborhood methods use a spatial
276 window or neighborhood surrounding the forecast and observed points and therefore
277 relax the condition for exact matches between forecasted and observed precipitation in
278 space, time and intensity.

279 Three of the single observation-neighborhood forecast approaches are used here
280 for evaluating the precipitation performance of the HIRLAM and the HARMONIE
281 simulations during the 15 sea breeze thunderstorm events ('observed but not forecast
282 episodes'). The three methods compare a neighborhood of forecast values to the
283 observation in the centre of the neighborhood. This mimics how a forecaster might
284 interpret high resolution model outputs when making a prediction for a particular
285 location. These particular methods were selected to assess the frequency of the forecasts
286 predicting appropriately high rainfall near the observed location, and to evaluate the
287 spatial scales over which a forecaster might consider the model output to be useful.

288 The first neighborhood technique corresponds to the 'multi-event contingency
289 table' proposed by Atger (2001), which measures the closeness of the forecasted
290 precipitation to the raingauge observations for multiple decision intensity, spatial and
291 temporal thresholds. Basically, the 'multi-event contingency table' builds on the
292 traditional 2x2 contingency table, in which hits, misses, false alarms, and correct
293 rejections are counted for forecasted and observed occurrences of an event (for
294 precipitation an event is usually defined as rain exceeding a given intensity threshold).
295 The method combines the customary intensity dimension with additional thresholds
296 such as spatial and temporal closeness. The method is based on the criterion that a
297 forecast is useful if the event is forecasted in at least one grid box within a specified
298 distance s of an observed event. If that is the case, then the forecast is counted as a hit; if

309 not, then the forecast is a miss. Forecasts for rain within a distance s of a location where
 300 rain was not observed are counted as false alarms.

301 The Hanssen and Kuipers (HK) score is computed in order to measure the
 302 success of the precipitation forecasts. The HK score (also known as true skill statistics -
 303 TSS- and Peirce's skill score -PSS-) represents a balance between the hit rate (H) and
 304 false alarm rate (F), i.e., $HK=H-F$, with a perfect value of 1:

$$305 \quad H = \frac{\text{hits}}{\text{hits} + \text{misses}} \quad (1) \quad F = \frac{\text{false alarms}}{\text{correct rejections} + \text{false alarms}} \quad (1)$$

306 The second neighborhood method applied here corresponds to the pragmatic
 307 technique defined by Theis et al. (2005) which uses the observed precipitation in the
 308 central grid box as a value to verify the model performance against the forecast
 309 precipitation probability within a neighborhood. A useful forecast has skill at detecting
 310 precipitation events and non-events, that is, it has high probabilities in the vicinity of
 311 observed events and low probabilities in the vicinity of observed non-events. The
 312 probabilistic verification metric corresponds to the Brier skill score (BSS) with respect
 313 to the sample climatology. The reference forecast (i.e., the forecast to beat) is the
 314 fraction of observations over the whole domain with rain meeting or exceeding the
 315 threshold.

$$316 \quad BSS = 1 - \frac{\frac{1}{N} \sum_N (\langle P_y \rangle_s - I_x)^2}{\frac{1}{N} \sum_N (\bar{P}_x - I_x)^2} \quad (2)$$

317 where N is the number of observation locations, I_x is an indicator (1=yes, 0=no) for an
 318 observed event at each location, $\langle P_y \rangle_s$ is the fraction coverage of forecast events within

319 the neighborhood of radius s , and \overline{P}_x is the domain mean value of I_x . BSS varies
 320 between minus infinity and 1, with larger values indicating more skillful forecasts.

321 The third and last neighborhood verification technique applied here is the
 322 conditional square root of ranked probability score (CSRR), defined by Germann and
 323 Zawadzki (2004). The definition of a useful forecast according to this metric is one with
 324 a high probability of matching the observed precipitation value. This method is based
 325 on the computation of the ranked probability score (RPS), which uses the frequencies of
 326 forecast precipitation values in logarithmically increasing intervals. The RPS quantifies
 327 the sum of squared differences in cumulative probability space for a multi-category
 328 probabilistic forecast (e.g., Wilks, 2011):

$$329 \quad RPS = \frac{1}{M-1} \sum_{m=1}^M (CDF_{y,m} - I_{x,m})^2 \quad (3)$$

330 where M is the number of forecast categories, $CDF_{y,m}$ is the cumulative probability of
 331 the forecast exceeding the threshold for category m , and $I_{x,m}$ is 1 if the observed rainfall
 332 meets or exceeds the threshold for category m and 0 if not. The square root of the RPS
 333 is the standard error of the forecast probability in probability space, and therefore values
 334 closer to 0 indicate better forecasts. In order to compare different rainfall episodes the
 335 square root of the RPS is normalized by the observed rain fraction.

$$336 \quad CSRR = \frac{\sqrt{RPS}}{P_{x>0}} \quad (4)$$

337 All three fuzzy methods were applied to the individual gridded forecasts. For
 338 each method and model-observation category, the results were aggregated over all cases

339 and plotted both as a function of neighborhood size and as quilt plots. The window sizes
340 were 1x1, 3x3, 5x5, up to 25x25, only plotting results for spatial scales from 2.5 up to
341 97.5-km. The rainfall thresholds used to define events and categories (needed by the
342 various scores) were 0.2, 1.0, 2.0, 5.0, 10.0, 20.0, and 50.0 mm, where an event is
343 defined as the rain equal to or exceeding the threshold. A more detailed description of
344 these three neighborhood verification techniques and metrics can be found in the review
345 article of Ebert (2008).

346 In section 3 we present neighborhood verification results for all the 15 sea
347 breeze thunderstorms and show an example of good high-resolution precipitation
348 forecast performance for an isolated sea breeze thunderstorm developed over the Iberian
349 System Mountains (ISM) on 7 August 2008.

350

351 **3. Results**

352 *3.1. Rain distribution from gridpoint match-ups*

353 The rain distribution plot displayed in Figure 2 shows that all three high-
354 resolution gridded rainfall forecasts overestimate the observed point precipitation
355 intensity distribution, with the HIR-D2.5 having a slightly greater over-prediction bias
356 than the coarser HIR-D5 model. The HAR-D2.5 did a much better job of representing
357 the observed rain amount distribution. The models also differed in their ability to
358 predict the maximum rainfall. For instance, forecast rain accumulations exceeding 20
359 mm are uncommon in the HIR-D5 (sample maximum of 22 mm), whereas the higher
360 resolution models produced much higher rain accumulations (sample maxima of 113
361 mm for HIR-D2.5 and 64 mm for HAR-D2.5). The observed maximum rain amount

362 was 90 mm (see Table 1). The overprediction bias has an impact on the neighborhood
363 scores discussed below.

364

365 *3.2. Multi-event contingency table method*

366 The multi-event contingency table technique, which checks whether a
367 precipitation forecast event is found near an observed value, suggest that the HIRLAM
368 model provided better forecasts than the HARMONIE model when low rain thresholds
369 were chosen, but for precipitation exceeding about 5 mm the HARMONIE model
370 performed more reasonably. This overall feature is clearly discernible looking at the
371 quilt plots shown in Figure 3 which display the HK metric for different rain intensity
372 thresholds (0.2, 1.0, 2.0, 5.0, 10.0, 20.0 and 50.0 mm; *x*-axes) as a function of various
373 window sizes (from 2.5 up to 97.5-km; *y*-axes). The magnitude of HK score is plotted
374 and the shade gives an idea of the model's performance, i.e., darker (orange colors) and
375 lighter (blue colors) shades represent better and poorer skills, respectively. In addition,
376 an intensity dependent skillful scale can be obtained by drawing a line through the
377 better score in the verification HK results as shown in Figure 3. Table 2 compares the
378 skillful scales and accompanying HK scores for three models. From the quilt plots and
379 table we can conclude that for rain thresholds of 5.0 mm and greater, the optimal scale
380 for the HARMONIE model is smaller than for the HIRLAM suites, meaning that one
381 needs to look over a smaller distance to find a forecast matching the observation,
382 without having too many false alarms for non-events. For instance, the 0.2 mm rain
383 threshold is best achieved at grid scale by the HIR-D2.5 (2.5 km; HK 0.34; Figure 3a)
384 and the HIR-D5 (5.0 km; HK 0.31; Figure 3b), and at a slightly higher scale but with
385 greater accuracy by the HAR-D2.5 (7.5 km; HK 0.37; Figure 3c). Higher horizontal grid

386 resolution for both the HIR-D2.5 and the HAR-D2.5 is crucial for obtaining smaller
387 optimal scales for rain intensity thresholds equal or greater than 1.0 mm in comparison
388 to the coarser HIR-D5. The optimal scales and HK scores found for the rain threshold of
389 50.0 mm are not representative since there were only 3 observations exceeding this
390 intensity.

391 Comparing the 2.5-km models the HAR-D2.5 model outperformed the HIR-
392 D2.5 model for most rainfall thresholds and scales, with HIR-D2.5 only scoring better
393 than HAR-D2.5 at the finest 2.5-km grid scale and large scales for 10-20 mm rain
394 thresholds. These results confirm the improved ability of the forthcoming non-
395 hydrostatic convection-permitting HARMONIE model suite as an operational NWP
396 model to capture isolated convection associated with mesoscale / local winds.

397

398 *3.3. Pragmatic method*

399 The BSS is used as a metric for the pragmatic neighborhood verification
400 technique in order to test how useful forecasts are in detecting events and non-events.
401 The reference forecast (i.e., the forecast to beat) is the fraction of observations in each
402 case that are events (i.e., the sample climatology over the whole domain). Figure 4
403 shows the quilt diagrams where the magnitude of the BSS metrics is plotted as a
404 function of the aforementioned intensity precipitation thresholds (x -axes) and the range
405 of spatial scales (y -axes). In general, the pragmatic method suggests that the forecasts
406 had poor skill for most rain rates and scales, with the HARMONIE scheme performing
407 best according to the BSS scores. To resolve this apparent contradiction in comparison
408 to the reasonable HK scores shown in Figure 3, it is important to recognize that both

409 neighborhood methods are addressing different questions; the pragmatic approach is a
410 much tougher test than the “at least one nearby forecast event”. Because all three
411 models over-predicted the number of rain events, it was difficult to beat the sample
412 climatology and the skill was negative for all scales and thresholds as shown in Figures
413 4a (HIR-D2.5), 4b (HIR-D5) and 4c (HAR-D2.5), and summarized in Table 3. We can
414 conclude that, according to this score, poorer performance occurred at small scales and
415 low to moderate thresholds. This skill according to the BSS metric was poorer for HIR-
416 D2.5 than for HIR-D5, while the HAR-D2.5 had the best performance due partly to its
417 more appropriate rain intensity distribution.

418

419 *3.4. Conditional square root of RPS*

420 The conditional square root of RPS method, which is an error metric with a
421 perfect value of 0, also showed the best performance for the HAR-D2.5, followed by the
422 HIR-D5 and the HIR-D2.5. As with the pragmatic method, the CSRR scores improve
423 when increasing to larger scales as shown in Figure 5. This method rewards forecasts
424 that have a peak in the (neighborhood-based) probability density function that is close to
425 the observed value. The threshold dependence is integrated into the overall score
426 through the forecast PDF. The forecast precipitation for the HIR-D2.5 tended to be
427 higher than for the HIR-D5 and the HAR-D2.5, which means that the forecast PDF
428 peaked even further away from the observation (see Fig. 2).

429

430 **4. Sea breeze front case study: 7 August 2008**

431 *4.1. Observational data*

432 The case study corresponds to a typical SBF that was not forecasted by the operational
433 HIRLAM numerical weather prediction and occurred over the Iberian system mountains
434 (SBCZ2) on 7 August 2008. We chosen this particular event because the highest
435 availability of grid observation points (i.e., 533 gauges; see Table 1) which helped for
436 better evaluating the performance of HIRLAM and HARMONIE forecasts. This non-
437 forecasted sea breeze thunderstorm produced severe convective activity with heavy
438 showers, large hail up to 3.5 cm in diameter, and gusty winds, causing significant
439 damage in some fruit groves and irrigated areas for vegetables, and also damaged
440 vehicles. The precipitation map displayed in Figure 6 confirms that the study case
441 corresponded to a very isolated sea breeze thunderstorm cell which accumulated
442 significant rainfall over a small area in the SBCZ2. For instance, 40 out of the 533 rain-
443 gauge points, i.e. 7.5%, measured precipitation ≥ 0.1 mm, whereas just 6 meteorological
444 stations reported rainfall ≥ 10.0 mm: Catí-Meteoclimatic (45.0 mm), Vallivana-CEAM
445 (27.8 mm), Morella-CHE (15.8 mm), Vallibona-CHJ (11.2 mm), Catí-CHJ (11.2 mm),
446 and El Boixar (10.0 mm).

447

448 *4.2. Operational weather forecast*

449 For the 7 August 2008 experiment the operational HIRLAM model used at the
450 AEMET was version 6.1.2 with a European domain of 0.16° (16-km length) to simulate
451 the large scale meteorological conditions, and an inner domain with fine resolution of
452 0.05° (5-km length) for the Iberian Peninsula and the Balearic Islands; 40 full sigma
453 levels in the vertical direction were used. The operational HIRLAM 6.1.2 version from
454 the AEMET partly failed in predicting this sea breeze thunderstorm, as is shown for

455 both the 0.16° (Figure 7a) and the 0.05° (Figure 7b) domains, where the 0.16° run gave
456 some light showers and the high resolution run did not produce any precipitation at all.

457

458 *4.3. HIRLAM and HARMONIE performance*

459 Figure 8 shows 24-h accumulated precipitation forecasts for the same
460 integration time from the HIR-D5 (Fig. 8a), the HIR-D2.5 (Fig. 8b), the HAR-D2.5
461 (Fig. 8c), and (d) the HAR-D2.5 (Fig. 8d) with a factor of 5 increase in precipitation
462 fall speed. Both the HIRLAM and the HARMONIE research simulations succeeded in
463 capturing the occurrence of convective showers over the SBCZ2 in comparison to the
464 operational HIRLAM model version 6.1.2. However, the HIR-D5 run predicted a weak
465 widespread precipitation signal and displaced far to the northwest; the HIR-D2.5 run
466 simulated heavier rainfall over a larger area located far to the north; and the HAR-D2.5
467 run forecasted the precipitation at the right place with a reasonable amount. For
468 instance, the precipitation maxima accumulated in 24 h was 6.6 mm for the HIR-D5,
469 92.5 mm for the HIR-D2.5, and 28 mm for the HAR-D2.5; the observed maximum was
470 45.0 mm in the Cati station. Forecast rain accumulations exceeding 45.0 mm were not
471 uncommon in the HIR-D2.5 model (seven grid-points sampled 24-h accumulated
472 precipitation greater than this threshold), whereas the maximum 24-h gridded rain
473 accumulations in the HIR-D5 model were less than 10 mm. In contrast, the HAR-D2.5
474 produced a reasonable number of grid-points with maximum rain accumulations similar
475 to the observations. Summarizing, the HIR-D2.5 version tended to strongly
476 overestimate precipitation at the rain-gauge points; the HIR-D5 version tended to
477 strongly underestimate rainfall at rain-gauge points; and the HAR-D2.5 version
478 performed better precipitation forecasts than both HIRLAM set-ups in terms of amount

479 and location, because it correctly reproduced the heavy rainfall along the sea breeze
480 boundary. We also tested the hypothesis that a large part of the overestimation of HIR-
481 D2.5 in comparison to HAR-D2.5 is associated with the fact that the fall speed is
482 infinity in the HIRLAM model suite, i.e., precipitation is put on the ground directly,
483 which is quite normal in hydrostatic models until recently. This was investigated by
484 increasing the fall speed of the rain, snow and graupel by a factor of 5 for the HAR-
485 D2.5 model suite. Fig. 8d shows a significant increase of precipitation by a factor of 2
486 or more over a large part of the precipitation covered area in Fig. 8c, with a maximum
487 precipitation of 55 mm whereas the original run produced a maximum precipitation
488 amount of 28 mm. This difference can be explained by the increase in fall speed as this
489 reduces the water loading of parcels in the updraft of the showers, increasing the
490 updraft wind speed and associated precipitation formation. Additionally, precipitation
491 spends less time in unsaturated air beneath the cloud, causing less evaporation of this
492 precipitation.

493 Examination of the wind field shown in Figure 9 reveals that the three model
494 setups resolved the development of a well-defined low-level convergence of sea
495 breezes (easterly-southeasterly flows from the Mediterranean sea) and synoptic flows
496 (westerly-southwesterly flows over the lower plateau of the IP). This convergence zone
497 played a major role in initiating severe weather on the afternoon of 7 August 2008.
498 However the location and strength of the low-level convergence of sea breezes is
499 determined by the model resolution, and this could hypothetically explain why the
500 position of the precipitation is much better in the HAR-D2.5 than in HIR-D5 and HIR-
501 D2.5.

502 The neighborhood verification metrics used here did not consider closeness in
503 time (Weusthoff, 2011), but time-series plots represent a way to look at the agreement
504 between HIRLAM and HARMONIE model results and the observations in terms of
505 timing. However, for this case-study significant precipitation amounts (i.e., ≥ 5.0 mm)
506 were available in only three AWS, which are shown in Figure 10. Here we compared
507 the hourly observed precipitation against the maximum modelled precipitation in an
508 area with radius of 5-km around the AWS. For the Vallivana-CEAM station (27.8 mm;
509 Fig. 9a) both the HIR-D5 and the HIR-D2.5 did not capture the hourly maxima of
510 precipitation that occurred at 15 UTC (15.0 mm) and 16 UTC (12.7 mm), whereas the
511 HAR-D2.5 correctly predicted the timing and also the amount (16.5 and 14.2 mm at 15
512 and 16 UTC, respectively). For the Morella-CHE station (15.8 mm; Fig. 9b), the hourly
513 maxima of precipitation occurring at 15 UTC (4.6 mm) and 16 UTC (11.2 mm) were
514 fairly reproduced by the HIR-D5 (2.3 mm at 16 UTC); the HIR-D2.5 detected two
515 peaks of precipitation at the time of this episode (4.7 and 3.4 mm at 15 and 16 UTC,
516 respectively) but incorrectly predicted precipitation during the rest of the day with a
517 maximum peak of rainfall at 22 UTC (6.5 mm). The HAR-D2.5 best represented this
518 episode, but with a maximum of 7.9 mm at 15 UTC, i.e., showing a timing error of one
519 hour before the observed maximum. Lastly, for the Michavilla-CHE station (8.2 mm;
520 Fig. 9c) again the HAR-D2.5 performed reasonably well for the hourly maximum of
521 precipitation observed at 15 UTC (5.8 mm against 7.6 mm for the HAR-D2.5); whereas
522 the HIR-D2.5 overestimated the amount and timing with a peak of maximum
523 precipitation of 73.4 mm at 17 UTC; and the HIR-D5 captured an hourly maximum of
524 4.5 mm but one hour later than the observed event. Therefore, we can conclude that, for
525 this particular case, the HARMONIE was also able to perform better in terms of timing
526 in comparison with the HIRLAM set-ups.

527

528 *4.4. Neighborhood verification of the HAR-D2.5 precipitation forecast*

529 Tartaglione et al. (2005) pointed out that single case-study verification by means
530 of non-parametric statistical methods may lead to unstable or poor results, due to the
531 paucity of the statistical sample. In addition, with high resolution models like the ones
532 applied here and with the operational forecasts increasing in resolution at the Weather
533 Services it is quite difficult to get good verification results, even when radar data are
534 available, due to the difficulty in matching between forecasts and observations.
535 However, here we show the neighborhood verification of the HAR-D2.5 as an example
536 of good forecast performance of the non-forecasted sea breeze thunderstorm that
537 occurred on 7 August 2008. The neighborhood verification of the HAR-D2.5
538 precipitation forecast against the maximum observed rain is shown in Figure 11. The
539 multi-event contingency table (Fig. 10a) displays strong HK scores for most rainfall
540 thresholds, and particularly for neighborhood sizes greater than 10-km. The pragmatic
541 approach (Fig. 10b) also reported positive BSS scores for most spatial scales and
542 rainfall thresholds. Lastly, the conditional square root of RPS method (Fig. 10c)
543 showed lower (better) than average values of CSRR. In general, the HARMONIE
544 forecast precipitation overlapped quite well with several observations of heavy rain,
545 demonstrating the ability of this model to reproduce isolated convection associated with
546 low-level convergence of sea breezes.

547

548 **5. Summary and discussion**

549

550 In this study we investigated the ability of both the three-dimensional
551 hydrostatic grid-point HIRLAM model (5.0-km and 2.5-km grid-spacing set-ups) and
552 the forthcoming operational non-hydrostatic spectral HARMONIE suite (2.5-km) in
553 predicting sea breeze thunderstorms that occurred over preferential convergence zones
554 in the eastern façade of the Iberian Peninsula, Spain. We simulated a set of 15 sea
555 breeze thunderstorms with intense convective activity that were not forecasted by the
556 operational HIRLAM system. Neighborhood or ‘fuzzy’ verification techniques were
557 applied to this set of high resolution regional forecasts making use of high-density
558 gridded raingauge measurements. The aim was to statistically compare the performance
559 of HIRLAM and HARMONIE suites with different grid resolutions in predicting this
560 dangerous sea breeze convective phenomena in the Mediterranean area of the Iberian
561 Peninsula.

562 The overall results revealed that both the more recent high resolution HIRLAM
563 and the HARMONIE model succeeded in capturing the occurrence of these convective
564 showers, which were missed in the operational HIRLAM runs, with the HARMONIE
565 model being most capable of delivering high quality forecasts of isolated convection
566 associated with sea breezes in terms of amount, location, and also time. To summarize,
567 it was found that the HIRLAM model at 5-km predicted light to moderate precipitation
568 signals (weak overprediction bias) and a bit displaced in space; the HIRLAM model at
569 2.5-km simulated excessive amounts of rainfall (strong overprediction bias) for large
570 areas, i.e., the precise location was not too accurate; whereas the HARMONIE suite at
571 2.5-km forecasted the precipitation in the right place at the right time with reasonable
572 amount. It was therefore found that the performance of HARMONIE was better for the
573 cases studied here as shown by the neighborhood verification metrics. Weusthoff et al.
574 (2010) also evaluated by means of ‘fuzzy’ techniques that high-resolution models give

575 better results for more localized convective precipitation events, and Van der Plas et al.
576 (2012) found that higher precipitation intensities were better represented in the
577 HARMONIE model.

578 The completely missed or strongly underestimated convection of the HIRLAM
579 versions and the experimental HIRLAM model at 5-km forecast are hypothetically
580 caused by the coarser resolution of these HIRLAM versions, combined with the less
581 developed physics in the operational AEMET runs of that time. Differences in
582 predicting isolated convection found between the coarser HIRLAM versions and the
583 highest resolution HIRLAM and HARMONIE set-ups can be linked to the impact of
584 model grid resolution and therefore the ability to skillfully resolve sea breezes and local
585 winds (Ries and Schlünzen, 2009), in combination with the absence of a deep
586 convection parameterization in these cases. Different horizontal grid spacing schemes
587 result in a different placement and strength of the sea breeze front since the temperature
588 gradient is only a few kilometers wide (Atkins and Wakimoto, 1997). In theory, finer
589 resolution allows much stronger horizontal gradients and updrafts at the sea breeze
590 front and a quicker development of convective showers, but sometimes even a 2.5-km
591 grid spacing is too coarse to simulate convection associated with sea breezes. Another
592 effect of the higher resolution is that orography gradients become steeper, having a
593 direct impact on resolving the development of upslope winds. Furthermore, we found
594 that the overestimation bias for the HIRLAM forecasts can be partly explained by the
595 fact that the precipitation fall speed is infinity in the HIRLAM model suite, i.e.,
596 precipitation is put on the ground directly, which is quite normal in hydrostatic models.

597 To conclude, the HARMONIE model showed more realistic forecasts of
598 precipitation associated with sea breeze convergence than those given by the HIRLAM

599 model, since it is better able to provide good estimates of the precipitation
600 accumulations and location of isolated convective events. This overall result confirms
601 the potential capability and skill of the forthcoming HARMONIE model as operational
602 for most European Weather Services to capture isolated showers and thunderstorms
603 associated with mesoscale / local winds in areas of complex terrain. This advance in
604 quantitative precipitation forecasting has practical applications for very short range
605 forecasting of local dangerous convective phenomena, which cause significant
606 economic damage and flooding in the Mediterranean area of the Iberian Peninsula
607 (Sánchez et al., 2003).

608

609 **Acknowledgements**

610 The authors wish to acknowledge the editor and two anonymous reviewers for
611 their detailed and helpful comments to the original manuscript. This research was
612 undertaken in the frame of the BEST/2010/014 (GV), the JAE-DOC043 (CSIC;
613 European Social Fund, FSE) and the JCI-2011-10263 grants. This research was
614 supported by the projects CGL2011-27574-C02-02 and CGL2011-27536/HID,
615 financed by the Spanish Commission of Science and Technology, and FEDER;
616 ACQWA (FP7-ENV- 2008-1-212250), financed by the European Commission; “Efecto
617 de los escenarios de cambio climático sobre la hidrología superficial y la gestión de
618 embalses del Pirineo Aragonés”, financed by “Obra Social La Caixa”; and CTP1/12
619 “Creación de un modelo de alta resolución espacial para cuantificar la esquiabilidad y
620 la afluencia turística en el Pirineo bajo distintos escenarios de cambio climático”,
621 financed by the “Comunidad de Trabajo de los Pirineos”. The authors would like to
622 thank the AEMET, CEAM, CHE, CHJ, IVIA, METEOCLIMATIC, SIAM, SIAR and

623 SMC for raingauge data; and to José-Antonio García-Moya and Estrella Gutiérrez-
624 Marco (AEMET) for providing the HIRLAM 6.1.2 outputs.

625

626 **References**

627

628 Aguilar, E., Auer, I., Brunet, M., Peterson, T.C., Wieringa, J., 2003. *Guidelines on*
629 *Climate Metadata and Homogenization*. World Meteorol. Organ., 52 pp. [Available
630 online at http://www.wmo.int/datastat/documents/WCDMP-53_1.pdf] (last accessed
631 1 November 2013)

632 Atger, F., 2001: Verification of intense precipitation forecasts from single models and
633 ensemble prediction systems. *Non. Proc. Geophy.* **8**, 401-417.

634 Atkins, N.T., Wakimoto, R.M., Weckwerth, T.M., 1995. Observations of the sea-breeze
635 front during CaPE. Part II: Dual-Doppler and aircraft analysis. *Mon. Wea. Rev.* **123**,
636 944-969.

637 Atkins, N.T., Wakimoto, R.M., 1997. Influences of the synoptic-scale flow on sea
638 breezes observed during CaPE. *Mon. Wea. Rev.* **125**, 2112-2130.

639 Azorin-Molina, C., Martin-Vide, J., 2007. Methodological approach to the study of the
640 daily persistence of the sea breeze in Alicante (Spain). *Atmosfera* **20**, 57-81.

641 Azorin-Molina, C., Guijarro, J.A., Baena-Calatrava, R., Jansà, A., 2007. Sea breeze
642 convergence and convective cloud frequencies from AVHRR data over Mallorca
643 Island. 7th EMS Annual Meeting/8th European Conference On Applications of

644 Meteorology, El Escorial, Madrid, Spain, EMS Annual Meeting Abstracts **4**,
645 EMS2007-A-00541

646 Azorin-Molina, C., Connell, B.H., Baena-Calatrava, R., 2009. Sea-breeze convergence
647 zones from AVHRR over the Iberian Mediterranean area and the Isle of Mallorca,
648 Spain. *J. Appl. Meteor. Climatol.* **48**, 2069-2085.

649 Baker, R.D., Lynn, B.H., Boone, A., Tao, W.-K., Simpson, J., 2001. The influence of
650 soil moisture, coastline curvature, and land-breeze circulations on sea-breeze
651 initiated precipitation. *J. Hydromet.* **2**, 193–211.

652 Bechtold, P., Pinty, J.-P., Mascart, P., 1991. A numerical investigation of the influence
653 of large-scale winds on sea breeze and land breeze type circulations. *J. Appl.*
654 *Meteor.* **30**, 1268-1279.

655 Blanchard D. O., López, R. E., 1985. Spatial patterns of convection in south Florida.
656 *Mon. Wea. Rev.* **113**, 1282–1299.

657 Calvo, J., 2007. Kain-Fritsch convection in HIRLAM. Present status and prospects.
658 *Hirlam Newsletter* **52**, 57-64.

659 Dailey, P. S., Fovell, R.G., 1999. Numerical simulation of the interaction between the
660 sea-breeze front and horizontal convective rolls. Part I: Offshore ambient flow.
661 *Mon. Wea. Rev.* **127**, 858-878.

662 de Bruijn, E. I. F., de Rooy, W. C., 2012. Evaluation of HARMONIE in the KNMI
663 parameterisation testbed. *Adv. Sci. Res.* **8**, 167-170.

664 Ebert, E.E., 2008. Fuzzy verification of high-resolution gridded forecasts: a review and
665 proposed framework. *Meteorol. Appl.* **15**, 51-64.

666 Fovell, R.G., Dailey, P. S., 2001. Numerical simulation of the interaction between the
667 sea-breeze front and horizontal convective rolls. Part II: Alongshore ambient flow.
668 *Mon. Wea. Rev.* **129**, 2057-2072.

669 Fovell, R.G., 2005. Convective initiation ahead of the sea-breeze front. *Mon. Wea. Rev.*
670 **133**, 264-278.

671 Germann, U., Zawadzki, I., 2004. Scale dependence of the predictability of precipitation
672 from continental radar images. Part II: probability forecasts. *J. Appl. Meteorol.* **43**,
673 74-89.

674 Gollvik, S., 2002. A snow model intended for HIRLAM. Proceedings of the
675 SRNWP/HIRLAM Workshop on Surface Processes, Turbulence and Mountain
676 Effects. Spanish Meteorological Agency, Madrid (Spain).

677 Gollvik, S., 2004. Surface modelling in Northern Europe. HIRLAM Baltic Workshop
678 Report, pp. 52-60.

679 González-Márquez, J., Heredia, M. A., 2001. Sea breeze convection in Mallorca.
680 Proceedings, V National Symposium of Forecasting, Madrid, Spain, Spanish
681 Meteorological Agency, pp. 1-6. (in Spanish)

682 Huang, X.-Y., Yang, X., Gustafsson, N., Mogensen, K., Lindskog, M., 2002. Four-
683 dimensional variational data assimilation for a limited area. HIRLAM Technical
684 Report **57**, 44 pp.

685 Ivarsson, K.I., 2007. The Rasch-Kristjansson large scale condensation; present status
686 and prospects. Hirlam Newsletter **52**, 50-56.

687 Kain, J.S., 2004. The Kain-Fritsch Convective Parameterization: An Update. J. Appl.
688 Meteor. **43**,170-181.

689 Kourzeneva, E., Samuelsson, P., Ganbat, G., Mironov, D., 2008. Implementation of
690 lake model FLake in HIRLAM. HIRLAM Newsletter **54**, 54-61.

691 Mazarakis, N., Kotroni, V., Lagouvardos, K., Argiriou, A.A., 2009. The sensitivity of
692 numerical forecasts to convective parameterization during the warm period and the
693 use of lightning data as an indicator for convective occurrence. Atmos. Res. **94**,
694 704–714.

695 Millán, M. M., Estrela, M. J., Miró, J., 2005. Rainfall Components: Variability and
696 spatial distribution in a Mediterranean area (Valencia region). J. Climate **18**, 2682-
697 2705.

698 Mironov, D. V., 2008. Parameterization of lakes in numerical weather prediction.
699 Description of a lake model. COSMO Technical Report, 11, Deutscher
700 Wetterdienst, Offenbach am Main, Germany, 41 pp.

701 Muppa, S.K., Anandan, V.K., Amit Kesarkar, K., Vijaya Bhaskara Rao, S., Narasimha
702 Reddy, P. 2012 Study on deep inland penetration of sea breeze over complex terrain
703 in the tropics. Atmos. Res. **104-105**, 209-216

704 Neumann, J., 1951. Land breezes and nocturnal thunderstorms. J. Meteor. **8**, 60-67.

705 Nicholls, M.E., Pielke, R.A., Cotton, W.R., 1991. A two-dimensional numerical
706 investigation of the interaction between sea breezes and deep convection over the
707 Florida Peninsula. *Mon. Wea. Rev.* **119**, 298-323.

708 Noilhan, J., Planton, S., 1989. A simple parameterization of land surface processes for
709 meteorological models. *Mon. Wea. Rev.* **117**, 536-549.

710 Papanastasiou, D.K., Melas, D., Lissaridis, I., 2010. Study of wind field under sea
711 breeze conditions; an application of WRF model. *Atmos. Res.* **98(1)**, 102-117.

712 Pérez-Landa, G., Ciais, P., Sanz, M.J., Gioli, B., Miglietta, F., Palau, J.L., Gangoiti, G.,
713 Millán, M.M., 2006. Mesoscale circulations over complex terrain in the Valencia
714 coastal region, Spain, Part 1: simulation of diurnal circulation regimes. *Atmos.*
715 *Chem. Phys. Discuss.* **6**, 2809-2852.

716 Petterssen, S., 1956. *Weather Analysis and Forecasting*, ed. McGraw-Hill Book
717 Company, New York.

718 Pielke, R.A., Song, A., Michaels, P.J., Lyons, W.A., Arritt, R.W., 1971. The
719 predictability of sea-breeze generated thunderstorms. *Atmosfera* **4**, 65-78.

720 Pielke, R.A., 1974. A three-dimensional numerical model of the sea breezes over South
721 Florida. *Mon. Wea. Rev.* **102**, 115-139.

722 Pielke, R.A., Cotton, W.R., 1977. A mesoscale analysis over south Florida for a high
723 rainfall event. *Mon. Wea. Rev.* **105**, 343-362.

724 Pielke, T.A., Mahrer, Y., 1978. Verification analysis of the University of Virginia three-
725 dimensional mesoscale model prediction over south Florida for July 1, 1973. *Mon.*
726 *Wea. Rev.* **106**, 1568-1589.

727 Purdom, J. F. W., 1976. Some uses of high-resolution GOES imagery in the mesoscale
728 forecasting of convection and its behavior. *Mon. Wea. Rev.* **104**, 1474-1483.

729 Ries, H., Heinke Schlünzen, K., 2009. Evaluation of a mesoscale model with different
730 surface parameterizations and vertical resolutions for the Bay of Valencia. *Mon.*
731 *Wea. Rev.* **137**, 2646-2661.

732 Rontu, L., Sattler, K., Sigg, R., 2002. Parametrization of subgrid-scale orography effects
733 in HIRLAM. Technical Report **56**, 46 pp.

734 Sánchez, J.L., Fernández, M.V., Fernández, J.T., Tuduri, E., Ramis, C., 2003. Analysis
735 of mesoscale convective systems with hail precipitation. *Atmos. Res.* **67-68**, 573-
736 588.

737 Savijärvi, H., 1990. Fast radiation parameterization schemes for mesoscale and short-
738 range forecast models. *J. Appl. Meteor.* **29**, 437-447.

739 Seity, Y., Brousseau, P., Malardel, S., Hello, G., Bénard, P., Bouttier, F., Lac, C.,
740 Masson, V., 2011. The AROME-France Convective-Scale Operational Model. *Mon.*
741 *Wea. Rev.* **139**, 976–991.

742 Simpson, J.E., 1994. Sea breeze and local wind, ed. Cambridge University Press, New
743 York.

744 Simpson, M., Warrior, H., Raman, S., Aswathanarayana, P.A., Mohanty, U.C., Suresh,
745 R., 2007. Sea-breeze-initiated rainfall over the east coast of India during the Indian
746 southwest monsoon. *Nat. Hazards* **42**, 401-413.

747 Strickler, M. W., 2003. *Sea-Breeze Fronts and Their Role in Convective Initiation*.
748 North Carolina State University, Raleigh, North Carolina, 22 pp.

749 Tartaglione, N., Mariani, S., Accadia, C., Speranza, A., Casaioli, M., 2005. Comparison
750 of rain gauge observations with modeled precipitation over Cyprus using
751 Contiguous Rain Area analysis. *Atmos. Chem. Phys.* **5**, 2147-2154.

752 Theis, S.E., Hense, A., Damrath, U., 2005. Probabilistic precipitation forecasts from a
753 deterministic model: a pragmatic approach. *Meteorol. Appl.* **12**, 257-268.

754 Tijm, A.B.C, Lenderink, G., 2003. Characteristics of CBR and STRACO versions.
755 *HIRLAM Newsletter* **43**, 115-124.

756 Undén, P., Rontu, L., Calvo, J., Cats, G., Cuxart, J., Eerola, K., Fortelius, C., Garcia-
757 Moya, J.A., Gustafsson, N., Jones, C., Järvenoja, S., Järvinen, H., Lynch, P.,
758 McDonald, A., McGrath, R., Navascues, B., Odegaard, V., Rodriguez, E.,
759 Rummukainen, M., Room, R., Sattler, K., Savijärvi, H., Sass, B.H., Schreur, B.W.,
760 The, H., Tijm, S., 2002. *The HIRLAM-5 scientific documentation*. HIRLAM-5
761 project. SMHI, S-60176, Norrköping, Sweden, 144 pp.

762 Van der Plas, E. V., Wichers Schreur, B., Kok, K., 2012. A quantitative evaluation of
763 the high resolution HARMONIE model for critical weather phenomena. *Adv. Sci.*
764 *Res.* **8**, 149-155.

765 Wakimoto, R. M., Atkins, N.T., 1994. Observations of the sea breeze front during
766 CAPE. Part I: Single Doppler, satellite, and cloud photogrammetry analysis. *Mon.*
767 *Wea. Rev.* **122**, 1092-1114.

768 Weusthoff, T., Ament, F., Arpagaus, M., Rotach, M.W., 2010. Assessing the Benefits of
769 Convection-Permitting Models by Neighborhood Verification: Examples from MAP
770 D-PHASE. *Mon. Wea. Rev.* **138**, 3418–3433.

771 Weusthoff, T., 2011. Neighbourhood verification in space and time. 5th Intl. Verification
772 Methods Workshop, 1-6 Dec. 2011, Melbourne, Australia.

773 Wilks, D.S., 2011. *Statistical Methods in the Atmospheric Sciences*. 3rd Edition.
774 Elsevier Inc.

775 Wilson, P.H., 2008. Warm-season lake-/sea-breeze severe weather in the Northeast.
776 Master thesis, Department of Earth and Atmospheric Sciences, University at
777 Albany, State University of New York, 115 pp.

778 Zhang, M., Lin, W., Bretherton, C.S., Hack, J.J., Rasch, P.J., 2003. A modified
779 formulation of fractional stratiform condensation rate in the NCAR community
780 atmospheric model (CAM2). *J. Geophys. Res. – Atmos.* **108**, 4035,
781 doi:10.1029/2002JD002523.

782

783 **List of Tables**

784

785 Table 1. Sea breeze thunderstorms selected for verifying the precipitation performance
 786 of the HIRLAM and the HARMONIE models suites. The number of raingauge points,
 787 the maximum of precipitation and location of this event are tabulated for each episode.

788

| Date | N° Raingauges | MaximumPrecip. (mm) | Station |
|------------|---------------|------------------------|-----------------|
| 31/08/2001 | 450 | 57.5 | Villena (A) |
| 19/04/2002 | 380 | 22.2 | Beneixama (A) |
| 29/09/2002 | 420 | 31.0 | Fontanars (V) |
| 18/06/2003 | 423 | 13.9 | Mosqueruela (T) |
| 25/08/2003 | 424 | 3.1 | Alcora (C) |
| 02/06/2004 | 361 | 12.0 | Talave (AB) |
| 20/06/2004 | 363 | 90.0 | Pina (C) |
| 21/06/2004 | 361 | 36.0 | Fredes (C) |
| 12/08/2005 | 363 | 1.7 | Ibi (A) |
| 28/09/2006 | 356 | 25.0 | Pinoso (A) |
| 08/10/2007 | 350 | 35.0 | Ibi (A) |
| 01/08/2008 | 323 | 34.0 | Villafranca (C) |
| 07/08/2008 | 533 | 45.0 | Cati (C) |
| 08/08/2008 | 325 | 53.7 | Relleu (A) |
| 11/09/2009 | 304 | 43.0 | Cati (C) |

789

790 (A) Alicante province, (AB) Albacete province, (C) Castellon province, (T) Teruel province, and (V)
 791 Valencia province. The (C) and (T) regions correspond to areas affected by the SBCZ2 (Iberian system
 792 mountains) and the (A) and (V) regions by the SBCZ3 (Prebetic mountain ranges)

793

794

795

796 Table 2. Optimal scale (in km) and the HK score found for all three NWP models as a
 797 function of the rain intensity thresholds (in mm).

| Model suite | | Rain threshold (in mm) | | | | | | |
|-------------|-------|------------------------|------|------|------|------|------|-------|
| | | 0.2 | 1.0 | 2.0 | 5.0 | 10.0 | 20.0 | 50.0* |
| HIR-D2.5 | Scale | 2.5 | 2.5 | 17.5 | 37.5 | 37.5 | 37.5 | 17.5 |
| | HK | 0.34 | 0.38 | 0.40 | 0.52 | 0.57 | 0.54 | 0.32 |
| HIR-D5 | Scale | 5.0 | 25.0 | 55.0 | 65.0 | 65.0 | 85.0 | --- |
| | HK | 0.31 | 0.40 | 0.45 | 0.49 | 0.35 | 0.16 | 0 |
| HAR-D2.5 | Scale | 7.5 | 22.5 | 22.5 | 27.5 | 22.5 | 27.5 | 37.5 |
| | HK | 0.37 | 0.45 | 0.45 | 0.53 | 0.59 | 0.59 | 0.63 |

798 *There were only 3 observations of 50 mm or more.

799

800

801

802

803

804

805

806

807

808

809

810 Table 3. As in Table 2 but for the BSS scores.

| Model suite | | Rain threshold (in mm) | | | | | | |
|-------------|-------|------------------------|-------|-------|-------|-------|-------|-------|
| | | 0.2 | 1.0 | 2.0 | 5.0 | 10.0 | 20.0 | 50.0* |
| HIR-D2.5 | Scale | 97.5 | 97.5 | 97.5 | 97.5 | 97.5 | 92.5 | 97.5 |
| | BSS | -5.11 | -2.52 | -1.69 | -1.16 | -0.59 | -0.16 | -0.03 |
| HIR-D5 | Scale | 95 | 95 | 95 | 95 | 95 | 95 | 65 |
| | BSS | -3.55 | -1.40 | -0.83 | -0.34 | -0.10 | -0.01 | 0 |
| HAR-D2.5 | Scale | 97.5 | 97.5 | 97.5 | 97.5 | 92.5 | 87.5 | 97.5 |
| | BSS | -0.48 | -0.18 | -0.12 | -0.15 | -0.10 | -0.02 | -0.03 |

811 *There were only 3 observations of 50 mm or more.

812

813

814

815

816

817

818

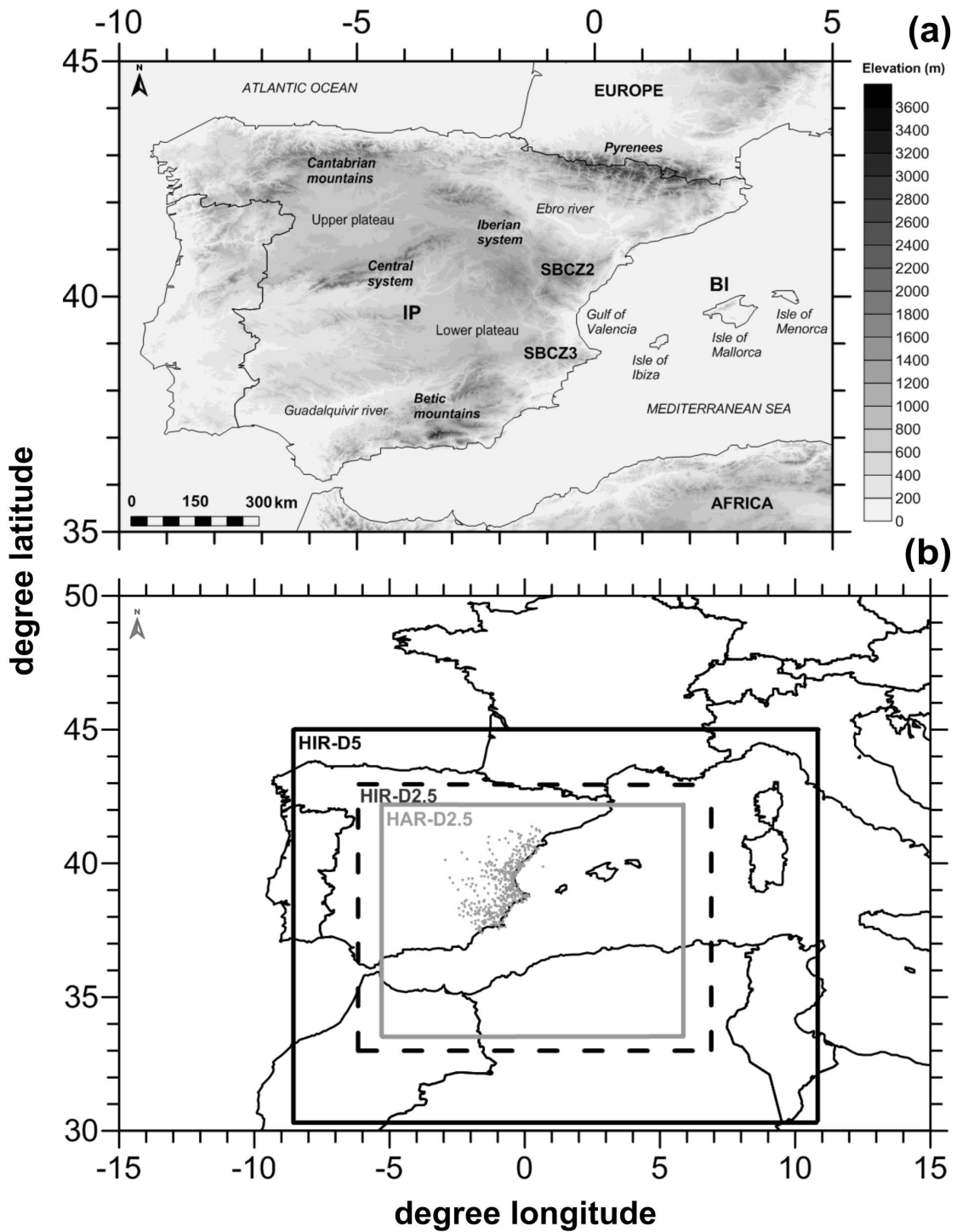
819

820

821

822

823



825

826 FIG. 1. (a) Terrain map of the Iberian Peninsula showing the complex topography of the
 827 study area focussed on the steep Iberian System mountains (SBCZ2) and the Prebetic
 828 mountains (SBCZ3). Specific locations cited to in the text and the most important

829 geographical features of the IP are shown on the map. (b) HIRLAM model
830 configuration showing a largest (HIR-D5; black solid line) and smallest (HIR-D2.5;
831 black dashed line) domain with horizontal grid lengths of 5.0-km and 2.5-km,
832 respectively, and also displaying the HARMONIE model configuration with a domain
833 of 2.5-km grid spacing (HAR-D2.5; grey solid line). The map shows location (grey
834 dots) of the high-density rain-gauge network for verification model output.

835

836

837

838

839

840

841

842

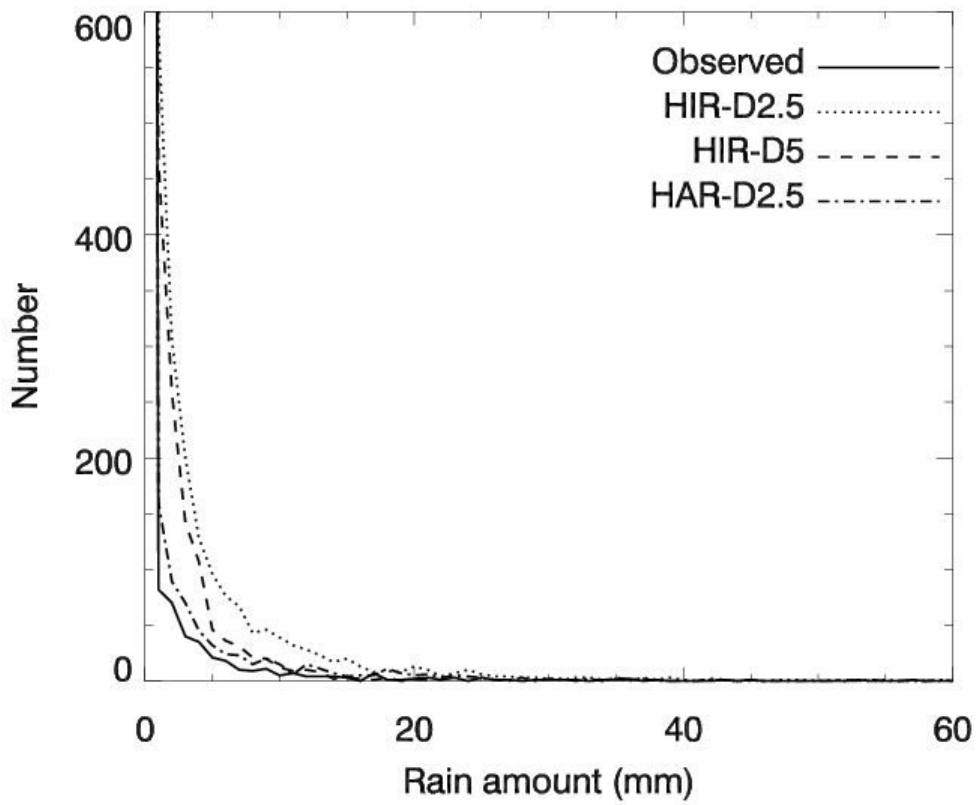
843

844

845

846

847



848

849 FIG. 2. Rain distribution from gridpoint match-ups using maximum observation in grid
850 box, pooled over all 15 non-forecasted sea breeze thunderstorms in the dataset.

851

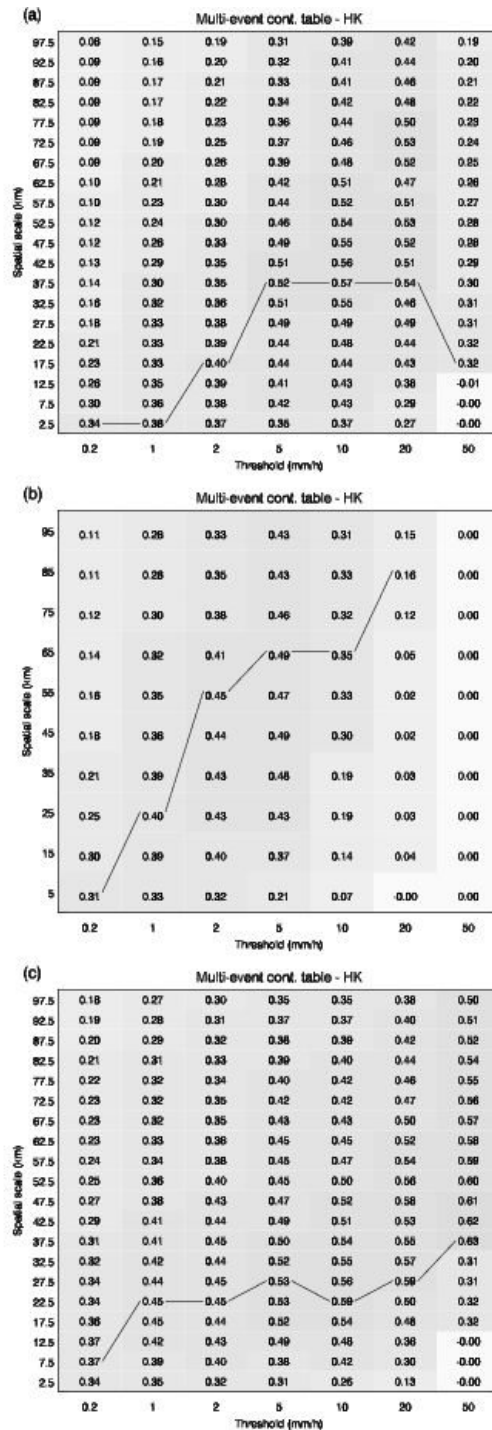
852

853

854

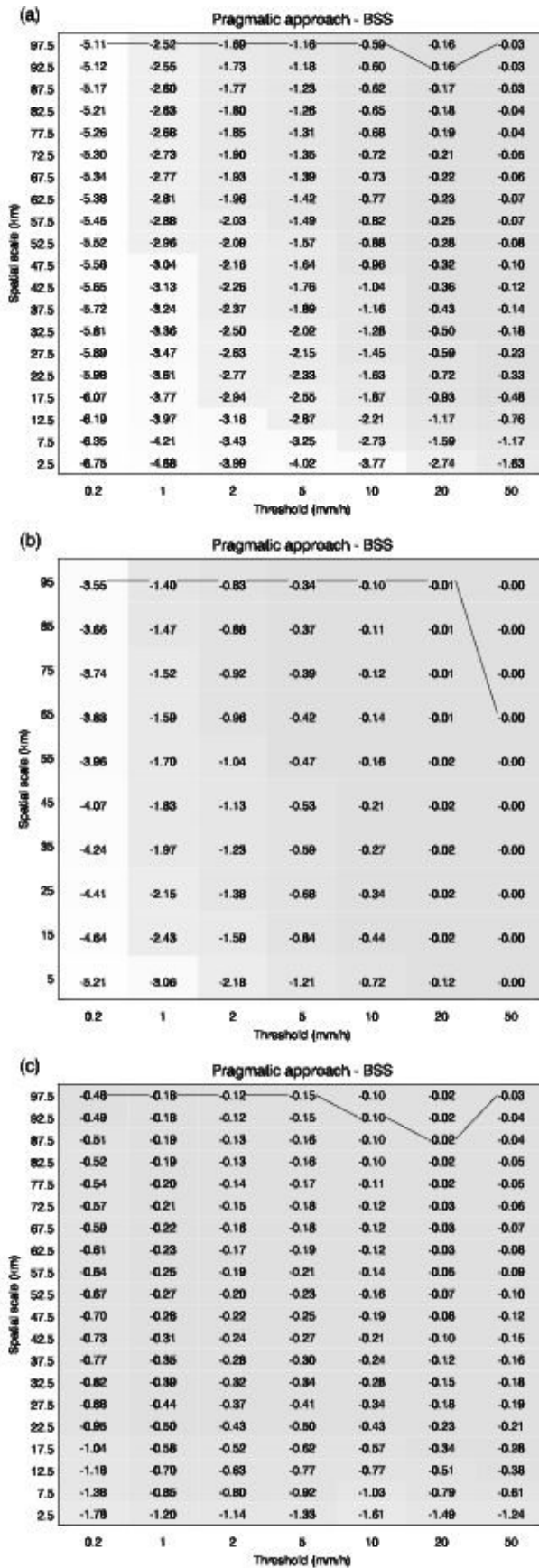
855

856



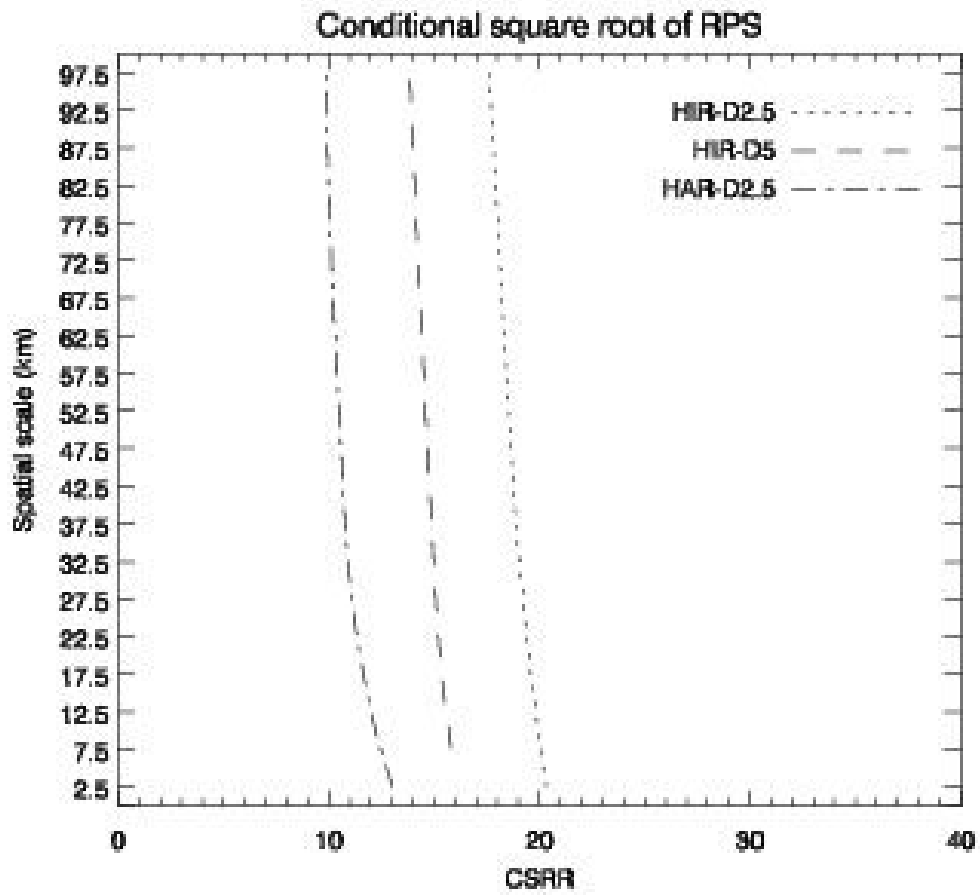
857

858 FIG. 3. Quilt plots of the HK scores as a function of different rain intensity threshold
 859 (0.2, 1.0, 2.0, 5.0, 10.0, 20.0 and 50.0 mm) and neighborhood sizes (from 2.5 up to
 860 97.5-km) for (a) the HIR-D2.5, (b) the HIR-D5 and (c) the HAR-D2.5. The darker and
 861 lighter shades show good or poor model performance, respectively. The lines indicate
 862 the better scale at each rain threshold.



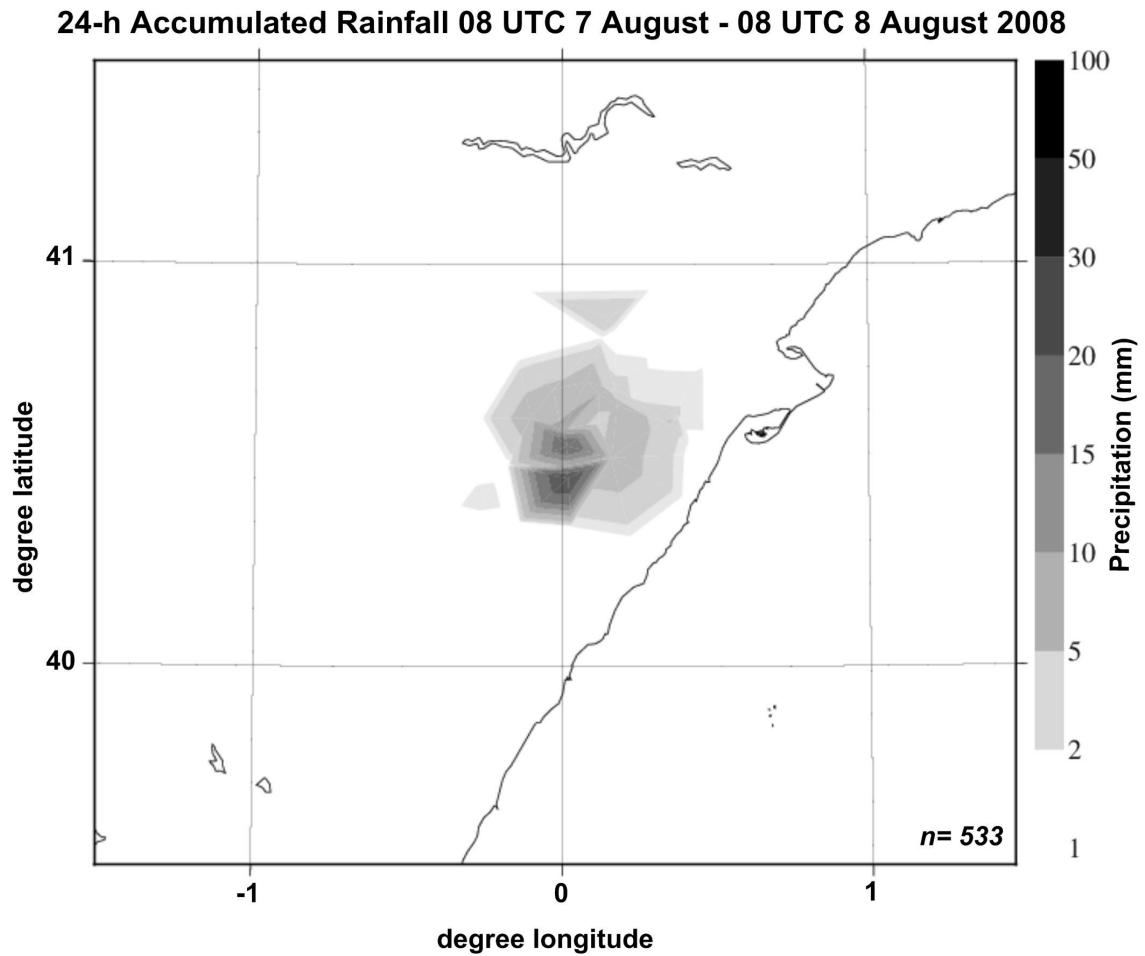
863

864 FIG. 4. As in Figure 3 but for the BSS computed for the pragmatic method.



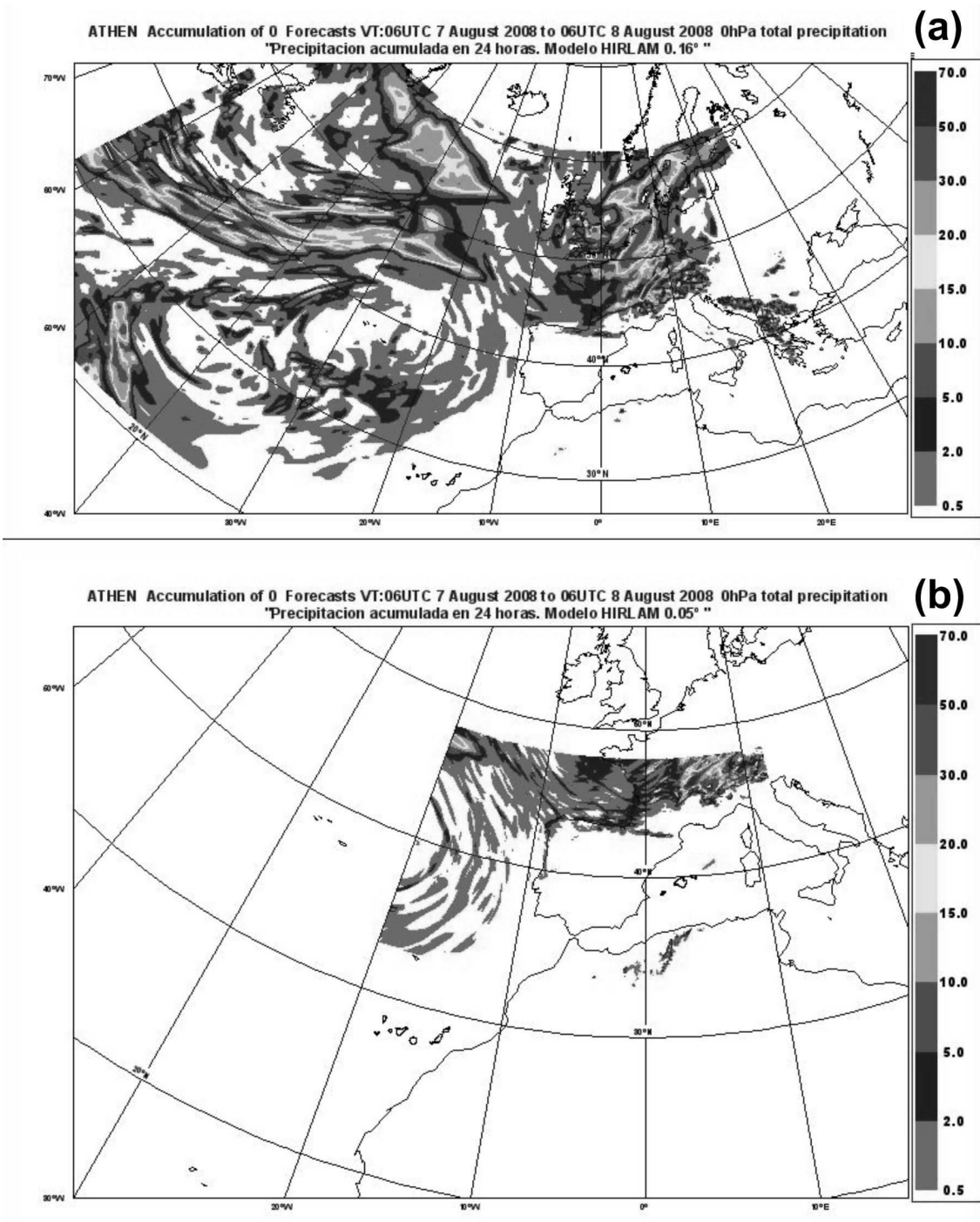
865

866 FIG. 5. Conditional square root of the RPS showing the CSRR metric as a function of
 867 different window sizes (from 2.5 up to 97.5-km).



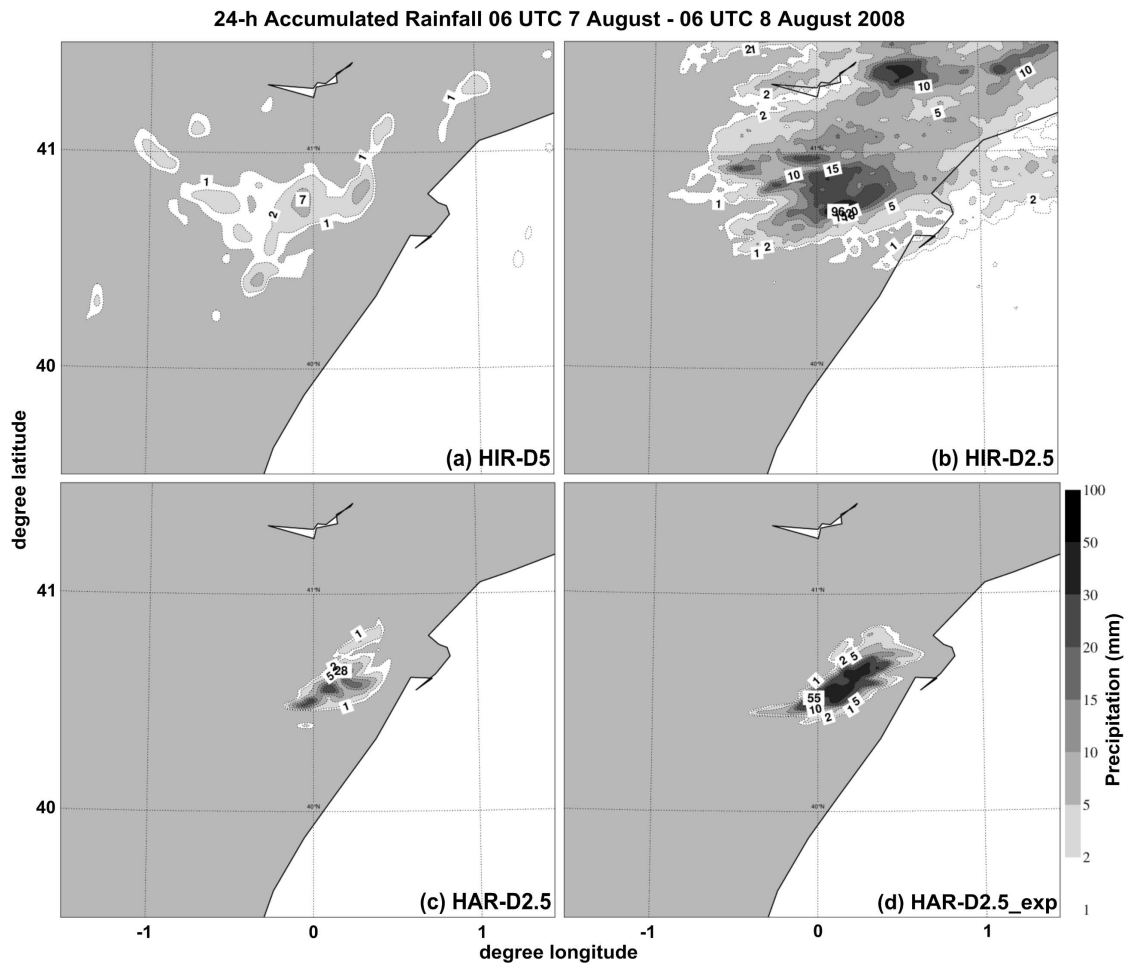
868

869 FIG. 6. Accumulated rainfall recorded between 08 UTC 7 August 2008 till 08 UTC 8
 870 August 2008. The high spatial density of measurements (n = sample size) was obtained
 871 by combining data from Hellmann (SYNOPS) and tipping bucket rain-gauges
 872 (automatic weather station; AWS). The map was produced using a Kriging
 873 interpolation.



874

875 FIG. 7. 24-h accumulated rainfall between 06 UTC 7 August 2008 till 06 UTC 8 August
 876 2008 simulated from (a) the HIR-D16 (0.16°, i.e., 16-km length) and (b) the HIR-D5
 877 (0.05°, i.e., 5-km length) operational HIRLAM 6.1.2 version from the AEMET.



878

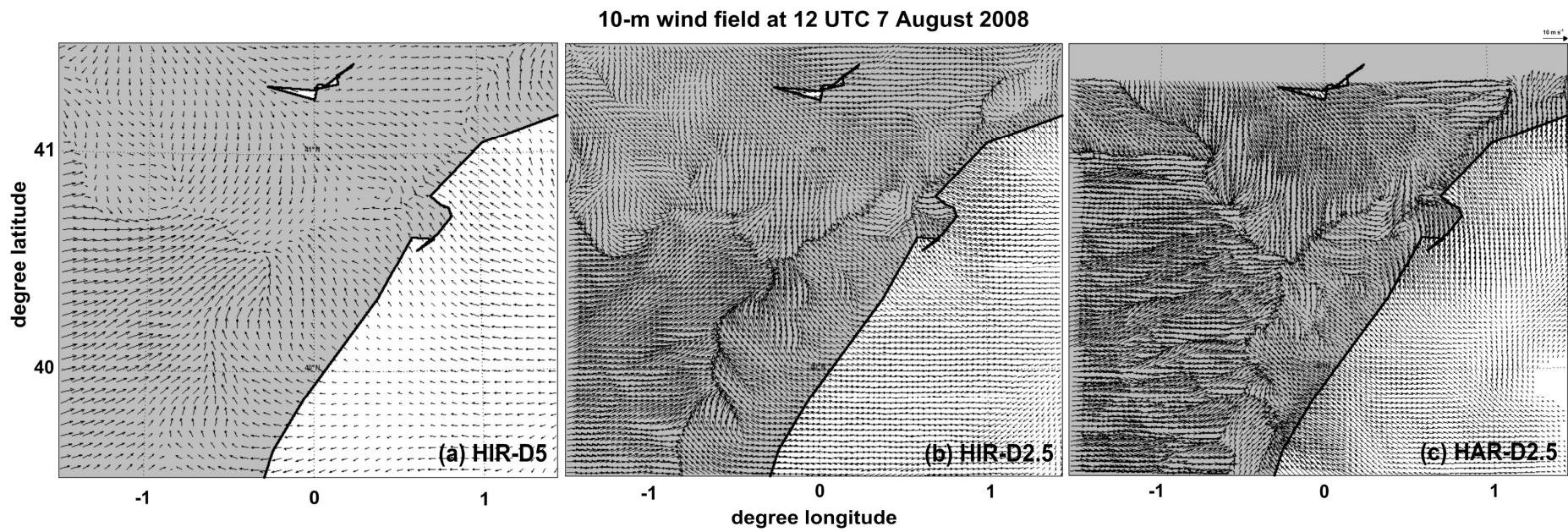
879 FIG. 8. 24-h accumulated rainfall between 06 UTC 7 August 2008 till 06 UTC 8 August
 880 2008 simulated from (a) the HIR-D5, (b) the HIR-D2.5, (c) the HAR-D2.5, and (d) the
 881 HAR-D2.5_exp with an increase in fall speed of precipitation by a factor of 5. Note that
 882 forecasted maps are plotted using the same boundaries and colors for comparison
 883 against the observed rainfall map shown in Figure 6.

884

885

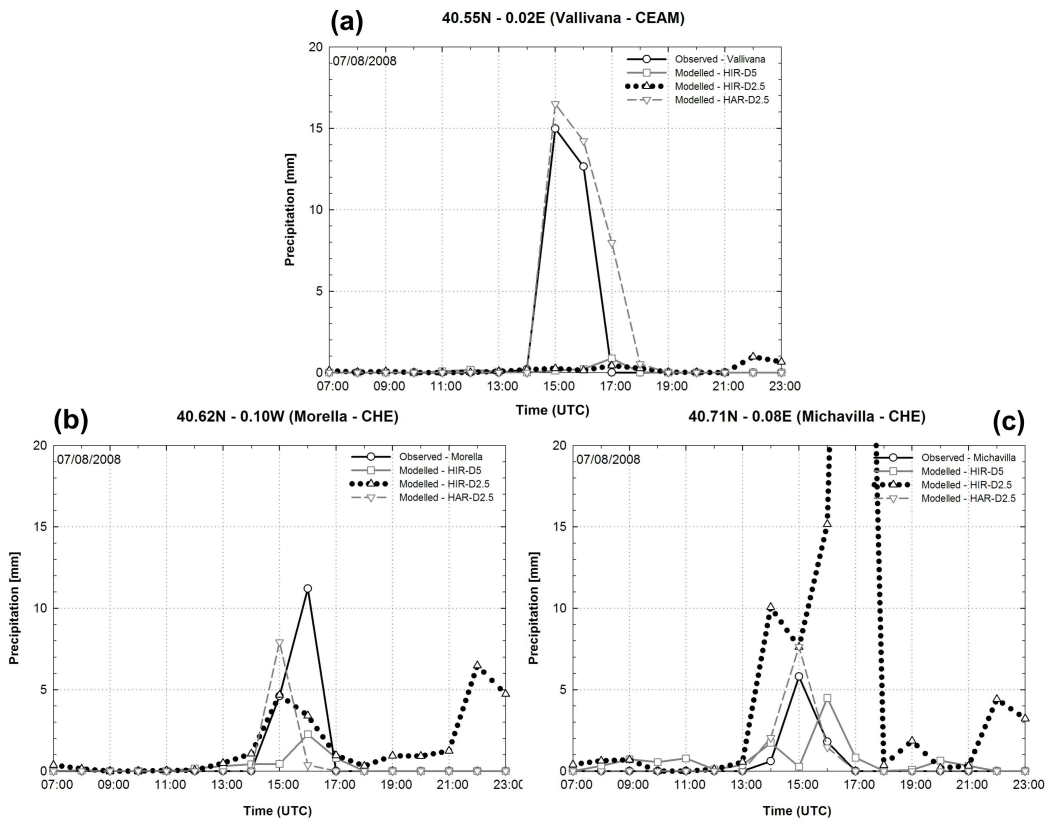
886

887



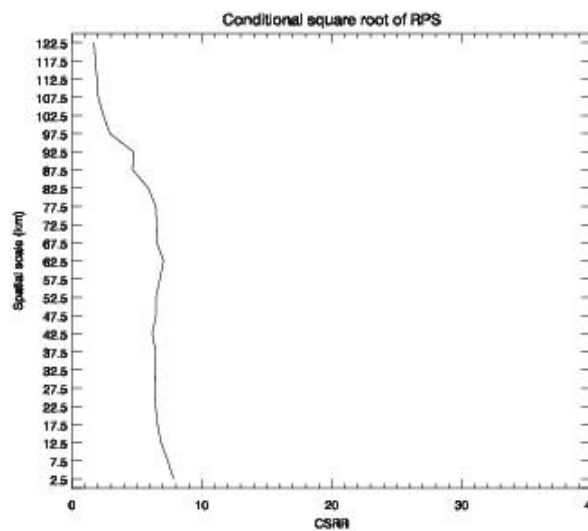
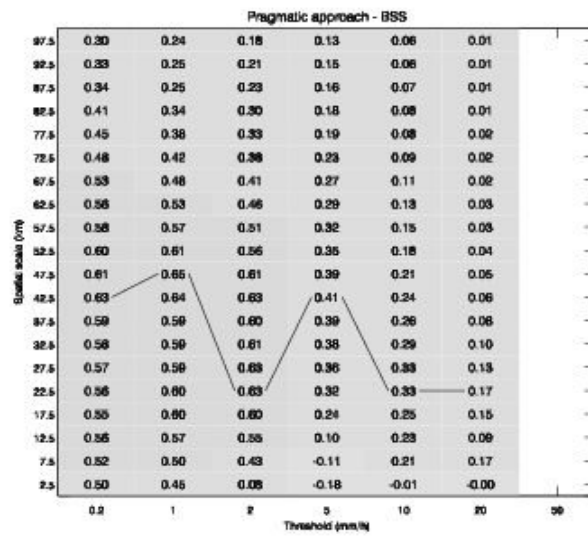
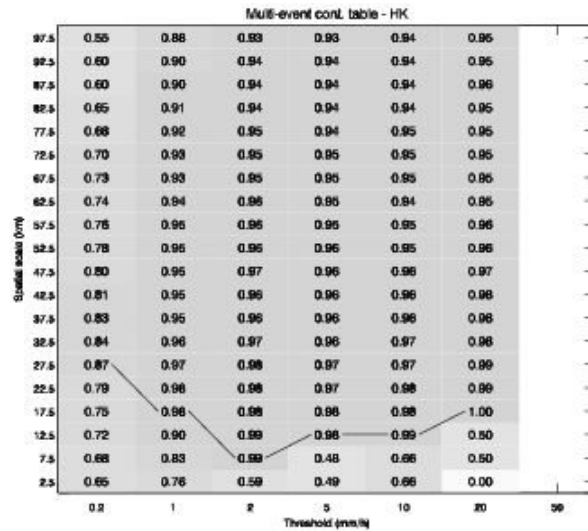
888

889 FIG. 9. 10-m wind vector simulated by (a) the HIR-D5, (b) the HIR-D2.5, and (c) the HAR-D2.5 at 12 UTC on 7 August 2008.



890

891 FIG. 10. Time series of observed and modelled precipitation in (a) Vallivana-CEAM,
 892 (b) Morella-CHE and (c) Michavilla-CHE from 0700 till 2300 UTC on 7 August 2008.
 893 The modelled values shown here correspond to the maximum precipitation in area with
 894 radius of 5-km around raingauge station. Note that modelled precipitation for HIR-D2.5
 895 exceeds the maximum y-axis for the Michavilla-CHE station.



896

897 FIG. 11. Neighborhood verification of the forecast performed by the HAR-D2.5 on 7
 898 August 2008 for (a) the HK scores and (b) the BSS metrics as a function of different

899 rain intensity threshold (0.2, 1.0, 2.0, 5.0, 10.0, 20.0 and 50.0 mm) and neighborhood
900 sizes (from 2.5 up to 97.5-km), and (c) the CSRR scores as a function of different
901 window sizes (from 2.5 up to 122.5-km). The quilt plots of the HK and the BSS metris
902 show in darker and lighter shades good or poor model performance, respectively. The
903 lines indicate the better scale at each rain threshold.

904

905

906

907

908

909

910