



1 The impact of lightning and radar data assimilation on the performance of very short term 2 rainfall forecast for two case studies in Italy

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14 Abstract

15 In this paper, we study the impact of t lightning and radar reflectivity factor data assimilation on
 16 the precipitation VSF (Very Short-term forecast, 3 hours in this study) for two relevant case
 17 studies occurred in Italy. The first case refers to a moderate localised rainfall over central Italy
 18 happened on 16 September 2017. The second case, occurred on 9 and 10 September 2017, was
 19 very intense causing damages in several geographical areas, especially in Livorno (Tuscany) where
 20 9 people lost their life.

21 The first case study was missed by several operational forecasts (from both public and private
 22 sectors), including that performed by the model used in this paper, while the Livorno case was
 23 partially predicted by operational models.

24 We use the RAMS@ISAC model (Regional Atmospheric Modelling System at Institute for
 25 Atmospheric Sciences and Climate of the Italian National Research Council), whose 3D-Var
 26 extension to the assimilation of RADAR reflectivity factor is shown in this paper.

27 Results for the two cases show that the assimilation of lightning and radar reflectivity factor,
 28 especially when used together, have a significant and positive impact on the precipitation
 29 forecast. The improvement compared to the control model, not assimilating lightning and radar
 30 reflectivity factor, is systematic because occurs for all the Very Short-term Forecast (VSF, 3h) of
 31 the events considered.

32 For specific time intervals, the data assimilation is of practical importance for civil protection
 33 purposes because it transforms a missed forecast of intense precipitation (> 40 mm/3h) in a
 34 correct forecast.

35 While there is an improvement of the rainfall VSF thanks to the lightning and radar reflectivity
 36 factor data assimilation, its usefulness is partially reduced by the increase of the false alarms in the
 37 forecast assimilating both types of data.

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39



40 1. Introduction

41 Initial conditions of numerical weather prediction (NWP) models are a key point for a good
 42 forecast (Stensrud and Fritsch, 1994; Alexander et al., 1999). Nowadays limited area models are
 43 operational at the resolution of few kilometres (< 5 km) and data assimilation of asyn local
 44 observations is crucial to correctly represent the state of the atmosphere at local scale (Weisman
 45 et al., 1997; Weygandt et al., 2008). This is especially important over the sea, where the absence
 46 of local observations and model deficiencies can misrepresent convection.

47 The assimilation of the radar reflectivity factor is important to improve the weather forecast
 48 because, being the reflectivity factor related to both the hydrometeors types and size, it can add
 49 information and eventually change the weather forecast. This is particularly important considering
 50 the high repetition rate (asynoptic data) and the high spatial resolution (local scale) of the radar
 51 data.

52 The first attempts to assimilate the radar reflectivity factor are reported in Sun and Crook (1997,
 53 1998), who expanded VDRAS (Variational Doppler Radar Analysis System) to include microphysical
 54 retrieval. Following these studies several systems to assimilate radar observations, both Doppler
 55 velocity and reflectivity factor, were developed (Xue et al., 2003; Zhao et al., 2006; Xu et al., 2010).
 56 All these studies showed the stability and robustness of assimilating radar observations as well as
 57 an improvement to the weather forecast.

58 Radar data are also assimilated in WRF (Weather Research and Forecasting, Skamarock et al.,
 59 2008; Barker et al., 2012) model both using 3DVar (Xiao et al., 2005, 2007; Barker et al., 2004) and
 60 4DVar approaches (Wang et al., 2013; Sun and Wang, 2012). The capability to assimilate radar
 61 data into WRF was recently applied to a heavy rainfall event over Central Italy by Maiello et al.
 62 (2014). They showed a notable and positive impact of the radar data assimilation on the
 63 precipitation forecast, also when radar data are assimilated together with conventional data.

64 In addition to those methods, which assimilate the radar reflectivity factor directly perturbing the
 65 hydrometeor contents predicted by the forecast models, there are indirect methods that aim at
 66 modifying other variables. In particular, the method proposed by Caumont et al. (2010) acts on the
 67 relative humidity field. It consists of two different steps: a 1D retrieval of relative humidity
 68 (pseudo-profile), which depends on the radar reflectivity factor observations, followed by 3D-Var
 69 assimilation of the pseudo-profile. This method, though less direct than perturbing hydrometeors,
 70 has the advantage to reduce the computational cost at the kilometeric scale, and to avoid
 71 questionable assumptions of the direct methods.



72 The choice of updating the moisture field directly is motivated by its greater impact on analyses
 73 and forecasts in comparison to that of hydrometeor-related quantities (e.g., Fabry and Sun, 2010).
 74 Caumont et al. (2010) showed that the method was able to improve the weather prediction for a
 75 case of heavy precipitation in southern France and for eight-day long assimilation cycle.
 76 The method was applied in other studies (Wattrelot et al., 2014; Ridal and Dalbom, 2016; Muller
 77 et al., 2017), or modified using 4D-Var in place of 3D-Var (Ikuta and Honda, 2011) showing in all
 78 cases its capability to improve the weather forecast. The method is also used in the operational
 79 context (Wattrelot et al., 2014).

80 Lightning are another important source of asynoptic data due to their ability to locate precisely
 81 the convection with few temporal gaps as well as, availability in real time thanks to the low
 82 bandwidth required for data transfer (Mansell et al., 2007). For these reasons, in the last two
 83 decades, there have been attempts to assimilate lightning into meteorological models both at low
 84 horizontal resolution, which need a cumulus parameterization scheme to simulate convection, and
 85 at convection permitting scales.

86 The first attempts to assimilate lightning in numerical weather prediction models (NWP) were
 87 based on relationships between lightning and rainfall rate estimated by microwave sensors on
 88 board polar satellites (Alexander et al., 1999; Chang et al., 2001; Jones and Macpherson, 1997). In
 89 this approach, the rainfall rate was computed as a function of lightning observations and then
 90 transformed into latent heat, which was assimilated. The results of these studies showed a
 91 positive impact of the lightning data assimilation on the forecast up to 24h also for fields at the
 92 large scale, as sea-level pressure, encouraging further researches.

93 Mansell et al. (2007) modified the Kain-Fritsch (Kain and Fritsch, 1993) cumulus convective scheme
 94 to force convection when/where flashes are observed while the convective scheme was not
 95 activated in the model simulation, demonstrating the potential of lightning to correctly locate the
 96 convection. A similar approach was introduced by Giannaros et al. (2016) into the WRF model
 97 showing the positive impact of the lightning data assimilation on the precipitation forecast up to
 98 24h for eight convective events occurred over Greece.

99 Fierro et al. (2012) introduced a methodology to assimilate lightning by modifying the water
 100 vapour mixing ratio simulated by the model according to a function depending on the flash-rate
 101 and on the simulated graupel mixing ratio. The water vapour could be assimilated by nudging
 102 (Fierro et al., 2012) or 3D-Var (Fierro et al., 2016).



Qie et al. (2014) extended the methodology of Fierro et al. (2012) to assimilate ice crystals, graupel and snow, showing promising results for deep convective events in China.

Federico et al. (2017a) implemented the methodology of Fierro et al. (2012) in RAMS@ISAC model, obtaining the systematic and significant improvement of the precipitation forecast at the very short range (3h) for twenty case studies occurred over Italy; the impact of lightning data assimilation for longer time ranges (6h-24h; Federico et al., 2017b) showed a considerable impact on the 6h precipitation forecast, with smaller (negligible) effects at 12 h (24 h).

In this paper we study the impact of the radar reflectivity factor and lightning observations data assimilation on the very short term (3h) rainfall prediction of two case studies over Italy. We use the method of Federico et al. (2017a) to assimilate lightning and the method of Caumont et al. (2010) to assimilate the radar reflectivity factor. The case studies occurred on September 2017. The first case, named Serano case, occurred on 16 September, was characterized by moderate-intense and localized rainfall. The second case, named Livorno case, occurred on 09-10 September was characterized by deep convection and very intense precipitation in several parts of Italy. Even if the Livorno case occurred before the Serano case, we will reverse the chronological occurrence in the discussion, ordering the event from the less intense to the most intense.

The Serano case was missed by the control forecast, not assimilating radar reflectivity factor and lightning. The Livorno event was partially predicted by the control forecast, which missed the abundant precipitation over Central Italy (see Section 4), and predicted the intense precipitation over Livorno delayed compared to the observations.

With respect to previous works, this study investigates the benefits brought by the combined use of radar and lightning observations into RAMS@ISAC, paving the way to systematic improvements of weather forecasts.

This paper is organized as follows: Section 2 gives details on the synoptic environment of the case studies showing daily precipitation, lightning and, for specific times, radar observations; Section 3 gives details on the meteorological model, lightning and radar data assimilation; Section 4 shows the results for five very short-term forecast (VSF), two for Serano and three for Livorno; Discussion and conclusions are given in Section 5.

131

132 **2. The case studies**

133 *2.1 The 16 September 2017 (Serano) case study*



134 During the 16 September 2017 the whole Italian country was under the influence of a cyclone that
 135 developed on the lee of the Alps. The storm crossed Italy from NW to SE leaving light precipitation
 136 over most of the peninsula with moderate rainfall over Central Italy. Figure 1 shows the
 137 precipitation recorded by the Italian raingauge network on 16 September 2017. A light
 138 precipitation (below 5 mm/day) is reported by 1018 raingauges out of the 1666 stations
 139 measuring precipitation (≥ 0.2 mm/day) on this day. Fourteen stations over Central Italy recorded
 140 more than 50 mm/day and 6 stations more than 60 mm/day. The maximum precipitation on this
 141 day was 90 mm/day in Città di Castello (Umbria Region). Because the meteorological radar closest
 142 to the maximum precipitation is over Serano mountain hereafter this event will be referred as
 143 Serano.

144 The synoptic condition during the event is shown in Figure 2. At 500 hPa (Figure 2a) a trough,
 145 elongated in the SW-NE direction, extends over Western Europe and air masses are advected from
 146 the SW towards the western Alps. The interaction between the airflow and the Alps generates a
 147 low pressure, at low levels, on the lee of the Alps over Northern Italy.

148 The situation at the surface (Figure 2b) shows the meteorological front represented by the
 149 equivalent potential temperature gradient between air masses advected over the Mediterranean
 150 Sea from NW and air masses advected from the South over the Tyrrhenian Sea, as a consequence
 151 of the pressure pattern that forms over the area. It is also notable the feeding of warm unstable
 152 air masses towards Central Italy.

153 Infrared satellite images (Figure 3), from 00 UTC on 16 September to 00 UTC on 17 September
 154 2017, show the cold front structure moving slowly from NW towards SE. Interestingly, at 00 UTC
 155 on 16 September, it is apparent a well-defined cloud system over Central Italy (red circle of Figure
 156 3a), which produced most of the daily precipitation observed between 43.50 and 45.0 N in the six-
 157 hours between 00 UTC and 06 UTC on 16 September 2017.

158 The well-defined cloud system over Central Italy is also clear in the radar Constant Altitude Plan
 159 Position Indicator (CAPPI) at 3 km above sea level at 02 UTC on 16 September (Figure 4). This
 160 CAPPI is formed by interpolating all the available data from the federated Italian radar network
 161 coordinated by the Department of Civil Protection (twenty-two radars, see Section 3.3 for their
 162 positions) and it is also referred as the national radar composite (hereafter also mosaic). Several
 163 convective cells exceeding 35 dBz can be noted on central-northern Italy. Importantly, the
 164 cloud system over Central Italy shown by the satellite infrared channel at 00 UTC (Figure 3a) and



165 that of the CAPPI at 02 UTC have a similar position, and the cloud system insisted for several hours
 166 over Central Italy (00-06 UTC).

167 Figure 5 shows the lightning recorded by the LINET network (Betz et al., 2009) on 16 September
 168 2017. More than 60.000 flashes were recorded for the whole day; most of them occurred during
 169 the afternoon and evening (the peak of more than 8000 flashes in one hour was at 22 UTC), but a
 170 secondary maximum occurred during the night on 16 September, from 00 UTC to 06 UTC. In this
 171 phase more than 2000 flashes were observed in Central Italy (see the green-blue dots in Figure 5).

172 From lightning observations, it follows that the storm had two main phases over Central Italy: the
 173 first one occurred during the night (00-06 UTC) and was characterised by the most intense rainfall;
 174 the second started after 18 UTC. In Section 4 one VSF for each phase will be considered.

175

176 2.2 The 09-10 September 2017 (Livorno) case study

177 During the days 09 and 10 September 2017, Italy was hit by a severe storm characterised by
 178 intense and widespread rainfall over the country. Damages to property were reported in several
 179 parts of Italy, while nine people died around Livorno, in Tuscany for causes related to the storm.

180 Figure 6a shows the precipitation on 09 September recorded by the Italian raingauge network.
 181 Rainfall was more intense over the Alps, where the maximum daily precipitation was observed
 182 (193 mm/day) and over Liguria, with precipitation of the order of 30-50 mm/day. One station over
 183 Tuscany reported 90 mm/day, showing that intense precipitation already started over the Region.
 184 The intensity of the storm on 09 September was high because 20 raingauges reported more than
 185 100 mm/day and 70 raingauges more than 60 mm/day, and, in most cases, this precipitation
 186 occurred in few hours. For example, the precipitation over Tuscany fell in the last 6 h of the day.

187 The following day (see Figure 6b) had higher rainfall. Precipitation occurred mainly over Central
 188 Italy, especially over Lazio, and over Northern Italy, in particular the North-East. In Tuscany, the
 189 two stations close to the sea, in the Livorno area, recorded about 150 mm/day mostly fallen in the
 190 hours between 00 and 06 UTC. The rainfall on 10 September was abundant: 256 stations out of
 191 2065 stations reported more than 60 mm/day, 60 of which recorded more than 100 mm/day.

192 Synoptic conditions leading to this storm are represented by the situation at 00 UTC on 10
 193 September, shown in Figure 7, when the storm was already producing precipitation over Northern
 194 Italy. At 500 hPa (Figure 7a) a trough extends from Northern Europe towards the Mediterranean.
 195 The interaction between air-masses and Western Alps generates a depression on the lee of the



196 Alps, over Northern Italy, which, in the following hours, crossed the whole peninsula from NW to
 197 SE. It is also noted the divergent flow over Central and Northern Italy favouring upward motions.
 198 At the surface, Figure 7b, it is apparent the equivalent temperature gradient over the western
 199 Mediterranean caused by the contrast between air masses pre-existing over the sea and air
 200 masses advected from France towards the Mediterranean Sea. The cyclonic circulation over the
 201 Ligurian Sea is forced by the low-pressure over the Northern Italy. The pressure field at the surface
 202 advects air masses from the South over the Tyrrhenian Sea. These air masses are unstable, i.e.
 203 humid and warm, and feed the cyclone during its development.
 204 From the synoptic point of view, this storm and the Serano case are similar and represent two
 205 cyclones developing on the lee of the Alps (Buzzi and Tibaldi, 1978). However, the Livorno case is
 206 more intense than Serano as shown by the larger rainfall, as well as by the more unstable air
 207 masses over the Tyrrhenian Sea that characterise the Livorno case.
 208 The notable intensity of the Livorno case is also confirmed by the lightning distribution (Figure 8).
 209 During the evening on 9 September (after 18 UTC) about 38.000 flashes were associated with the
 210 propagation of the storm from NW to SE. On 10 September about 170.000 flashes were recorded
 211 along Italy, following the movement of the storm propagating from NW to SE. So, more than
 212 200.000 flashes were recorded from 18 UTC on 09 September to 00 UTC on 11 September, which
 213 are more than twice those recorded for Serano.
 214 Satellite images (thermal infrared channel, 10.8 micron; Figure 9) show the extension of the cloud
 215 coverage every 12 hours. It is well evident the cloud system associated with the cold front that
 216 extends over Europe and moves from north-west to south-east. More specifically, the satellite
 217 image at 00 UTC shows the cloud system over Livorno area (red circle in Figure 9b), before the
 218 main precipitation event over Tuscany (00-06 UTC), while Figure 9c shows the cloud system over
 219 Central Italy (orange circle), at the end of the period of intense precipitation over Lazio (06-12
 220 UTC).
 221 We conclude the synoptic analysis of the case study with two CAPPI at 3 km observed by the radar
 222 network of the Department of Civil Protection. The CAPPI of Figure 10a, at 00 UTC on 10
 223 September, shows the cloud system over Tuscany with reflectivity factor up to 40 dBz. Other
 224 clouds are producing rainfall over northern Italy. The cloud system remained stationary over
 225 Tuscany for the period 00-06 UTC, with new cells developing over the sea and moving towards the
 226 land for six hours, causing the flood in Livorno. The CAPPI of Figure 10a is the last assimilated by
 227 the VSF of 00-03 UTC on 10 September shown in Section 4.



Figure 10b shows the CAPPI of the national radar mosaic at 3 km above the sea level and at 06 UTC. The cloud system is moving towards Central Italy with reflectivity up to 45 dBz. Other cloud systems are apparent over northern Italy. Figures 10a-10b well represents the movement of the storm towards SE and Figure 10b shows the last CAPPI assimilated by the 06-09 UTC VSF shown in Section 4.

233

234 **3.Data and Methods**


235 *3.1 RAMS@ISAC and simulations set-up*

The RAMS@ISAC is used for the numerical experiments of this work. The model is based on the RAMS 6.0 model (Cotton et al., 2003) with the addition of four main features, as well as a number of minor improvements. First, it implements additional single moment microphysical schemes, whose performance is shown in Federico (2016): among them, the WSM6 (Hong and Lim, 2006) is used in this paper. Second, it predicts the occurrence of lightning following the methodology of Dahl et al. (2012), and the implementation and performance of the scheme is discussed in Federico et al. (2014). Third, the model can assimilate lightning through nudging (Fierro et al., 2012; Federico et al., 2017a). Forth, the model implements a 3D-Var data assimilation system (Federico, 2013, hereafter also RAMS-3DVar), whose extension to the radar reflectivity factor is shown in this paper (Section 3.3).

The list of the main physical parameterisation schemes used in the simulations of the RAMS@ISAC discussed in this paper is shown in Table 1.

Considering the domains and the configuration of the grids (Figure 11 and Table 2), two different set-ups are used for Serano and Livorno. For the first case, we use the domains D1 and D2, while for Livorno we use also the third domain D3. The first domain covers a large part of Europe and extends over the North Africa. For this domain, the horizontal resolution of the grid is 10 km (R10). The second domain extends over the whole Italy and part of Europe and the grid has 4 km horizontal resolution (R4). The third domain covers the Tuscany Region, has 4/3 km horizontal resolution (R1), and it is used for Livorno to represent with more detail the precipitation field.

The resolutions and the extensions of the grids in the vertical direction are the same for the three domains and cover the troposphere and the lower stratosphere.

The nesting between the first and second domains is one-way, while the nesting between the second and the third domains is two-ways 



259 The VSF is implemented as shown in Figure 12. First a run using the R10 configuration is
 260 performed using the GFS analysis/forecast cycle issued at 12 UTC as initial and boundary
 261 conditions. This run, which starts at 12 UTC on 16 September for Serano and at 12 UTC on 09
 262 September for Livorno, lasts 36 h and doesn't assimilate radar reflectivity factor or lightning.
 263 Starting from 12 UTC, ten VSF are performed using R4 (for Serano) or both R4 and R1 (for Livorno).
 264 The VSF lasts 9h and uses R10 simulation as initial and boundary conditions (one-way nesting). The
 265 9h forecast is divided into two parts: the first six hours are the assimilation stage when the
 266 assimilated source of observations are continuously used to constrain the VSF to the observations
 267 whereas the last three hours are dedicated to the forecast stage when the VSF freely evolve
 268 without external constrains. During the assimilation stage, flashes are assimilated with the
 269 nudging technique (Section 3.2), while radar reflectivity is assimilated every one-hour by the
 270 Caumont et al. (2010) method (Section 3.3).

271 It is noted that data assimilation is performed in the domain D2 (R4) only and the innovations are
 272 transferred to the domain D3 (R1), for the Livorno case, by the two way-nesting.

273 The verification of the VSF for precipitation is done by visual comparison of the model output with
 274 the raingauge network of the Department of Civil Protection, which has more than 3000
 275 raingauges all over Italy (Davolio et al., 2015).

276 In addition we consider the FBIAS (Frequency Bias; range $[0, +\infty)$), where 1 is the perfect score,
 277 i.e. when no misses and false alarms occur), POD (Probability of Detection; range $[0, 1]$, where 1 is
 278 the perfect score and 0 the worst value), ETS (Equitable Threat Score; range $[-1/3, 1]$, where 1 is
 279 the perfect score and 0 is a useless forecast) and HR (Hit Rate or correct proportion; range $[0, 1]$,
 280 where 1 is the perfect score and 0 the worst value) computed from 2x2 dichotomous contingency
 281 tables (Wilks, 2006) for different rainfall thresholds (0.2 mm/3h, 1mm/3h and from 2mm/3h to
 282 the maximum thresholds, i.e. 40 mm/3h for Serano and 60 mm/3h for Livorno, every 2 mm/3h). In
 283 particular, defining the hits (a , a hit occurs when both the precipitation forecast and the
 284 corresponding raingauge observation are above or equal to a rainfall threshold), false alarms (b , a
 285 false alarm occurs when the precipitation forecast is above or equal to a rainfall threshold, while
 286 the corresponding raingauge observation is below the threshold); misses (c , a missing occurs when
 287 the forecast precipitation is below a rainfall threshold, while the corresponding raingauge
 288 observation is above or equal to the threshold), a correct no forecast occurs when both the
 289 precipitation forecast and the corresponding observation are below a rainfall threshold), we have:



$$\begin{aligned}
 FBIAS &= \frac{a+b}{a+c} \\
 POD &= \frac{a}{a+c} \\
 ETS &= \frac{a-a_r}{a+b+c-a_r}; \quad a_r = \frac{(a+b)(a+c)}{a+b+c+d} \\
 HR &= \frac{a+d}{a+b+c+d}
 \end{aligned} \quad (1)$$

where a_r is the probability to have a correct forecast by chance (Wilks, 2006). The hits, false alarms, misses and correct no forecast are computed comparing the precipitation forecast at four RAMS@ISAC grid points surrounding a raingauge and taking among them the closest value to the raingauge measurement (nearest-neighbour). In this way, we tolerate a spatial error of $D*(2)^{1/2}$ for the rainfall forecast, where D is the model grid spacing (4 km or 4/3 km depending by the case considered). Because the scores are computed for the second and third RAMS@ISAC domains, we tolerate spatial errors of 5.7 km and 1.9 km, respectively.

3.2 Lightning data assimilation

The lightning data assimilation scheme, introduced in previous papers (Federico et al., 2017a; 2017b), is shown here for completeness.

The method starts by computing the water vapour mixing ratio q_v :

$$q_v = Aq_s + Bq_s \tanh(CX)(1 - \tanh(Dq_s^\alpha)) \quad (2)$$

Where coefficients are set to $A=0.86$, $B=0.15$, $C=0.30$, $D=0.25$, $\alpha=2.2$, q_s is the saturation mixing ratio at the model atmospheric temperature, and q_g the graupel mixing ratio (g kg^{-1}). X is the number of flashes falling in a grid box of domain D2 (R4) in the past five minutes. The mixing ratio q_v of Eq. (2) is computed only for grid points where flashes are recorded, i.e. X is greater than zero. More specifically, for each grid point we consider the number of flashes falling in a grid box centred at the grid point in the last five minutes. The mixing ratio of Eqn. (2) is compared with that predicted by the model. If the mixing ratio of Eqn. (2) is larger than the simulated one, the latter is changed with the value given by Eqn. (2), otherwise the modelled mixing ratio is left unchanged. This method can only add water vapour to the forecast.

The check and eventual substitution of the water vapor is performed every five minutes and it is made only in the charging zone (-10°C , -25°C).

Lightning data are provided by the LINET network, which has more than 500 sensors over worldwide with the greatest density over Europe.



318

319 *3.3 Radar data assimilation*

320 The method assimilates radar CAPPI that are operationally provided by the Italian Department of
 321 Civil Protection (DPC). Radar data are provided over a regular Cartesian grid with 1 km horizontal
 322 resolution and for three vertical levels (2, 3, 5 km above the sea level) and radar observations can
 323 be considered as vertical profiles. These CAPPIs at the three different altitudes of 2, 3, and 5km
 324 can be considered as under-sampled vertical profiles. CAPPIs are composed starting from the 22
 325 radars of the Italian Radar Network (Figure 13) 19 operating at the C-band (i.e., 5.6 GHz) and 3 at
 326 X-band (i.e., 9.37 GHz). Data quality control and CAPPI composition is performed by DPC and no
 327 additional quality control is applied in this paper. Before entering the data assimilation, the
 328 Cartesian grid is reduced to 5 by 5 km by choosing one point every five of the Cartesian grid
 329 provided by DPC in order to reduce the dimensionality of the problem and to account, at least in
 330 part, for the correlation error of the observations.

331 The methodology to assimilate radar reflectivity factor is that of Caumont et al. (2010), named
 332 1D+3DVar, which is a two-step process: first, using a Bayesian approach inspired to GPROF (Olson
 333 et al., 1996; Kummerow et al., 2001), 1D pseudo-profiles of model variables are computed, then
 334 those pseudo-profiles are assimilated by 3DVar. Both steps are discussed shortly.

335 The first step computes a pseudo-profile of relative humidity weighting the model profiles of
 336 relative humidity around the radar profile (Bayesian approach). In particular:

$$337 \quad \mathbf{z}_o^p = \frac{\sum_i \mathbf{RH}_i W_i}{\sum_j W_j} \quad (3)$$

338 Where \mathbf{RH}_i is the RAMS@ISAC vertical profile of relative humidity at a grid point inside a square of
 339 $50 \times 50 \text{ km}^2$ centred at the radar vertical profile, W_i is the weight of each profile and \mathbf{z}_o^p is the
 340 relative humidity pseudo-profile. The summation is taken over all the grid points inside a square of
 341 $50 \times 50 \text{ km}^2$ around the observed profile and the denominator is a normalisation factor. The
 342 weights are determined considering the agreement between the simulated and observed
 343 reflectivity factor:

$$344 \quad W_i = \exp \left\{ -\frac{1}{2} \left[\mathbf{z}_o - h_z(x_i) \right]^T \mathbf{R}_z^{-1} \left[\mathbf{z}_o - h_z(x_i) \right] \right\} \quad (4)$$

345 Where h_z is the forward observation operator, transforming the background column \mathbf{x}_i into the
 346 observed reflectivity factor. The forward observation operator is specific for the WSM6



microphysics scheme and is available in WRF release 3.8. It assumes Marshall-Palmer hydrometeors size-distribution, Rayleigh scattering, and depends on the mixing ratios of rain, graupel and snow.

The observation error matrix \mathbf{R}_z in Eqn. (4) is assumed diagonal, i.e. observation errors are uncorrelated, and its value is $n\sigma^2$, where σ is 1 dBz and n is the number of available observations in the vertical profile (from 1 to 3).

It is important to note that the method is not able to force convection when the model has no rain, snow and graupel in a square around (50*50 km²) a specific radar profile with reflectivity factor greater than zero. In this case, the pseudo profile of relative humidity is assumed saturated above the condensation level and with no data below to force convection into the model.

It is also noted that the method is able to dry the model when the reflectivity factor is simulated but not observed, by giving more weight to the drier relative humidity profiles simulated by RAMS@ISAC in Eqn. (3).

The pseudo-profiles computed with the procedure introduced above, are then used as observations in the RAMS-3DVar data assimilation (Federico, 2013), minimising the cost-function:

$$J(\mathbf{x}) = \frac{1}{2}(\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}_b) + \frac{1}{2}(\mathbf{y}_o - h(\mathbf{x}))^T \mathbf{B}^{-1}(\mathbf{y}_o - h(\mathbf{x})) \quad (5)$$

Where \mathbf{x} is the state vector giving the analysis when J is minimized, \mathbf{x}_b is the background, \mathbf{B} and \mathbf{R} are the background and observations error matrices, \mathbf{y}_o is the observation vector and h is the forward observation operator transforming the state vector into observations. The cost function in RAMS-3DVar is implemented in incremental form (Courtier et al., 1994; see Federico 2013 for the details).

The background error matrix is computed using the NMC method (Parrish and Derber, 1992; Barker et al. 2004) applied to the HyMeX-SOP1 hydrological cycle in the Mediterranean Experiment – First Special Observing Period occurred in the period 6 September-6 November 2012; Ducroq et al., 2014), which has been chosen because it contains several heavy precipitation events over Italy (Ferretti et al., 2014).

In the RAMS-3DVar, the background error matrix is divided in three components along the three spatial directions (x , y , z). The \mathbf{B}_x and \mathbf{B}_y matrices take into account for the spatial correlation of the background error. They are assumed Gaussian with length-scales between 20 and 30 km, depending on the vertical level. Again, these distances are computed using the NMC methods (Barker et al., 2012).



378 The B_z matrix contains the error for the water vapor mixing ratio, which is the control variable
 379 used in RAMS-3DVar. This error is about 2 g/kg at the surface and decreases with height. In
 380 particular, it is larger than 0.5 g/kg below 4 km, and less than 0.2 g/kg above 5 km.

381 It is noted that cross correlations among variables are neglected in this study and the applications
 382 of the RAMS-3DVar affects the water vapor mixing ratio only.

383 Because the lightning data assimilation perturbs the water vapor mixing ratio, it follows that the
 384 data assimilation presented in this study changes only this parameter.

385

386 4. Results

387 4.1 Serano

388 In this section we analyse two VSF forecasts of the Serano case. The first period (03-06 UTC) is the
 389 most intense, while the second period (18-21 UTC) corresponds to a rejuvenating phase of the
 390 storm.

391

392 4.1.1 Serano: 03-06 UTC 16 November 2017

393 In this period, an intense and localised storm hit the central Italy, while light precipitation occurred
 394 over northern Italy (Figure 14a). Considering the storm over central Italy, 10 raingauges observed
 395 more than 30 mm/3h, 6 more than 40 mm/3h, 3 more than 50 mm/3h and 1 more than 60
 396 mm/3h, the maximum observed value being 63 mm/3h.

397 The CTRL forecast, Figure 14b, misses the storm over central Italy and underestimates
 398 considerably precipitation over Northern Italy, giving unsatisfactory results.

399 The assimilation of the radar reflectivity factor improves the forecast, as shown by Figure 14c. In
 400 particular, the RAD forecast shows localized precipitation (30-35 mm/3h) close to the area where the
 401 most abundant precipitation was observed. However, the maximum precipitation is
 402 underestimated. Another interesting improvement of the RAD forecast compared to CTRL is the
 403 precipitation over northern Italy, whose area is much more in agreement with observations
 404 compared to CTRL.

405 The precipitation forecast given by LIGHT, Figure 14d, shows some improvements compared to
 406 CTRL because the precipitation over central Italy has a maximum of 25-30 mm/3h, close to the
 407 area where the maximum precipitation was observed. LIGHT, however, has a worse performance
 408 compared to RAD because it misses the light precipitation over northern Italy. Also, similarly to
 409 RAD, LIGHT underestimates the maximum precipitation.



410 RA forecast, Figure 14e, shows the best performance. The precipitation over central Italy is
 411 better represented because the maximum rainfall (40-45 mm/3h) is in reasonable agreement with
 412 observations, and also because the area with intense precipitation (> 25 mm/3h) is elongated in
 413 the SW-NE direction in agreement with raingauge measurements, giving a much better idea of the
 414 real storm intensity compared to RAD and LIGHT, as well as CTRL. The light precipitation over
 415 northern Italy is well represented by RADLI.

416 Figure 14f shows the POD, computed for the domain of Figure 14a, for the time period considered.
 417 CTRL and LIGHT show a poor forecast compared to RAD and RADLI, underlining the importance of
 418 the assimilation of reflectivity factor observations for this phase of the storm. The POD of RADLI is
 419 0.33 for the 30 mm/3h threshold (3 stations out of 10 where correctly predicted). This represents
 420 a good performance considering that the intense precipitation is localized and we used the
 421 nearest neighbour methodology to compute the score, which, for the specific grid resolution,
 422 limits to 5.7 km the displacement error.

423 Figure 14f also shows the significant improvement of RAD and RADLI for the light rainfall forecast
 424 because the POD for the 0.2 mm/3h threshold increases from 0.5 of CTRL (0.55 for LIGHT) to
 425 about 0.85 for both RAD and RADLI.

426 The ETS score shows again the positive impact of the data assimilation, especially radar reflectivity
 427 factor, on the rainfall forecast for this phase of the storm, the best performance given by RADLI.

428 The proportion of correct forecast, Figure 14h, is larger than 84% for all configurations. HR,
 429 however, is lower for RAD and RADLI compared to other configurations because of the larger
 430 number of false alarms given by the assimilation of radar reflectivity factor.

431 It is finally remarked that lightning and reflectivity factor data assimilation acted synergistically
 432 because the simulation assimilating both data performs much better than the simulations
 433 assimilating only one kind of observation, either radar reflectivity factor or lightning.

434

435 4.1.2 Serano: 18-21 UTC 16 September 2017

436 In this phase, rainfall occurred mainly over central Italy with moderate-heavy amounts. In
 437 particular, 51 raingauges measured more than 10 mm/3h, 13 more than 20 mm/3h, 3 more than
 438 30 mm/3h and 2 between 40 mm/3h and 50 mm/3h (Figure 15a). Rainfall was also observed over
 439 north-western Italy with 12 raingauges observing more than 10 mm/3h, 7 more than 20 mm/3h, 4
 440 more than 30 mm/3h, and 3 between 40 mm/3h and 50 mm/3h.



441 The CTRL forecast, Figure 15b, shows little precipitation over central Italy, giving an unsatisfactory
 442 forecast, while the forecast over north-western Italy is well represented even if displaced few tens
 443 of kilometres to the North of the real occurrence.

444 The RAD forecast, Figure 15c, is better than CTRL. Firstly, the rainfall pattern over central Italy is
 445 well predicted but the maximum values are underestimated; secondly, the rainfall over north-
 446 western Italy is simulated more to the South compared to CTRL, more in agreement with
 447 observations.

448 The LIGHT forecast, Figure 15d, improves considerably the rainfall prediction over central Italy and
 449 the maximum values forecast. LIGHT are more in agreement with observations compared to
 450 RAD. The rainfall over north-western Italy is shifted to the North compared to raingauges
 451 measurements, similarly to CTRL.

452 RADLI forecast, Figure 15e, shares features with both RAD and LIGHT. For example, the maximum
 453 values over central Italy are similar to LIGHT, while the rainfall over north-western Italy is similar
 454 to RAD.

455 It is also noted that the precipitation over central Italy for thresholds higher than 20 mm/3h covers
 456 a wider area compared to LIGHT, extending towards the SW, giving a better representation of the
 457 observed precipitation.

458 RADLI has the best POD among the simulations, ranging from 0.8 (0.2 mm/3h) to 0.2 (38 mm/3h),
 459 followed by LIGHT and RAD. The improvement given by data assimilation to the CTRL forecast is
 460 notable for all experiments assimilating data (radar reflectivity factor and/or lightning). The ETS
 461 score shows that RADLI and LIGHT forecasts are useful up to about 40 mm/3h, while RAD forecast
 462 has a lower performance. Again, ETS shows a significant improvement of the simulations with data
 463 assimilation compared to CTRL. Despite the higher POD, RADLI has a lower ETS than LIGHT. This
 464 behaviour is found also for other VSF and is caused by the larger number of false alarms in the
 465 RADLI forecast, especially when compared to LIGHT.

466 The proportion of correct forecast (Figure 15h) is larger than 75% for all thresholds. The larger
 467 number of false alarms given by RADLI and RAD compared to LIGHT is notable in the lower values
 468 of HR for the configurations assimilating radar reflectivity factor.

469 Despite the bigger number of false alarms, the forecast given by RADLI is the best among all
 470 forecasts, because it clearly shows the moderate precipitation occurring over central and north-
 471 western Italy.

472




473 4.2 Livorno

474 The Livorno case lasted for several hours starting at 18 UTC on 9 September 2017 and ending
 475 more than a day later. The most intense phase in Livorno and its surroundings was observed
 476 during the night between 9 and 10 September. In the following, we will show three representative
 477 VSF (3h), including the most intense phase in Livorno.

478

479 4.2.1 Livorno: 18-21 UTC 9 September 2017

480 During this period, the precipitation started to hit intensely  Livorno and its surroundings (point A
 481 in the Figure 16a). Figure 16a shows the rainfall observed between 18 and 21 UTC on 9
 482 September. Over Tuscany there are three stations around Livorno (the yellow-red raingauges of
 483 Figure 16a close to label A) reporting more than 30 mm/3h: 31 mm/3h, 37 mm/3h, and 55
 484 mm/3h, respectively. The precipitation is spread over Liguria, Tuscany and Emilia Romagna, with
 485 130 raingauges, of the 517 raingauges available in this time interval, measuring more than 10 mm
 486 in 3h and 25 raingauges measuring more than 20 mm in 3h.

487 CTRL forecast for this period is shown in Figure 16b. It is apparent that CTRL misses the
 488 precipitation over coastal Tuscany, while that over the Apennines (label B in Figure 16a) is
 489 underestimated. CTRL predicts the precipitation over Liguria but the amount is overestimated
 490 being the forecast amount over 75 mm/3h for some stations, while observations have maximum
 491 values between 25 mm/3h and 30 mm/3h.

492 Figure 16c shows the RAD precipitation forecast. The impact of the reflectivity factor data
 493 assimilation is notable. Compared to CTRL, the precipitation covers a larger area and reaches the
 494 coastal part of Tuscany. The precipitation over Livorno is not well predicted, the amount being 10-
 495 15 mm/3h.

496 The precipitation over Liguria is still overestimated by RAD but to a lower extent compared to
 497 CTRL. The assimilation of radar reflectivity factor can increase or decrease the water vapour
 498 content of the simulations, depending on the reflectivity factor observed and simulated, and the
 499 lower precipitation over Liguria for RAD compared to CTRL is an effect of the reduction of water
 500 vapor caused by the data assimilation of radar reflectivity factor.

501 Figure 16d shows the precipitation simulated by LIGHT. The rainfall reaches the Tuscany coast and
 502 extends more to the East in the northern part of the domain compared to both CTRL and RAD,
 503 being LIGHT more in agreement with observations.



504 The precipitation forecast around Livorno is 25-30 mm/3h, however the precipitation over the sea,
 505 30 km far from the location of the most intense precipitation observed in the 3h, reaches 45 mm
 506 in 3h, giving the hint of a potentially intense storm.

507 The rainfall over Liguria is overestimated by LIGHT, but to a less extent compared to both RAD and
 508 CTRL. The LIGHT simulation moves the storm southeastward faster than other configurations,
 509 leaving less rain over Liguria compared to RAD and CTRL.

510 Figure 16e shows the rainfall of RADLI forecast. The precipitation field shares some characteristics
 511 with the LIGHT simulation and some others with RAD simulation. For example, the precipitation in
 512 the northern part of the domain, similarly to LIGHT, extends more to the East compared to RAD.
 513 The precipitation swath over Tuscany coast is similar to that of RAD, but shifted southward. The
 514 maximum precipitation in the Livorno area is 20-25 mm /3h, underestimating the observed
 515 maximum precipitation, but being closer to the observed position compared to RAD. The
 516 precipitation over Liguria is overestimated by RADLI.

517 POD score, Figure 16f, shows that CTRL performance is improved by data assimilation. LIGHT
 518 performs better than RAD up to 16 mm/3h, while RAD performs better for larger thresholds,
 519 thanks to a better simulation of the precipitation over the Apennines (label B of Figure 16a).
 520 Interestingly, the POD of RADLI follows LIGHT up to 16 mm/3h and RAD for larger thresholds,
 521 having, overall, the best score.

522 ETS score, Figure 16g, confirms the results of POD. CTRL forecast is useful ($ETS > 0$) up to
 523 14mm/3h, LIGHT up to 18 mm/3h, RAD and RADLI up to 22 mm/3h.

524 HR is lower for CTRL up to 22 mm/3h because it has a lower number of hits. For larger thresholds
 525 RADLI has the lowest HR because of its higher number of false alarms compared to other
 526 configurations.

527 In summary, for the period of the onset of high precipitation over Livorno, the assimilation of
 528 lighting or radar reflectivity factor or both data improves the precipitation forecast giving hint of
 529 intense precipitation in the Livorno area for both LIGHT and RADLI simulations. However, the
 530 maximum precipitation in Livorno is underestimated by the VSF forecast even with data
 531 assimilation, while the precipitation over Liguria is overestimated.

532

533 4.2.2 Livorno: 00-03 UTC 10 September 2017

534 This period represents the most intense phase of the storm in Livorno. In particular, the raingauge
 535 close to the label A (Figure 17a) reported 151 mm/3h (Collesalveti), while the one close to the



label B measured 82 mm/3h. Among the 518 raingauges reporting valid data, 75 observed more than 10 mm/3h, 31 more than 20 mm/3h, 17 more than 30 mm/3h, 9 more than 40 mm/3h, and 6 more than 50 (also 60) mm/3h.

The CTRL precipitation forecast is shown in Figure 17b. The forecast is poor because it misses the precipitation swath from the coast towards NE. Indeed, a precipitation swath is forecasted about 50 km to the North of the real occurrence, but it is less wide compared to the observations.

The forecast of RAD, Figure 17c, shows that the assimilation of radar reflectivity factor gives a clear improvement to the forecast. The largest precipitation in the coastal part of the swath (we searched the maximum value in the area with longitudes between 10.20E and 10.70E and latitudes between 43.10N and 43.60N) is 94 mm/3h, clearly showing the occurrence of a heavy precipitation event. Another local maximum is shown in the southern part of the domain (label B). The maximum location is well represented, but the forecast value is 55 mm/3h while the observed maximum is 82 mm/3h.

An improvement, compared to both CTRL and RAD, is given by the assimilation of lightning (Figure 17d). Also for this simulation there is a precipitation swath from coastal Tuscany to the Apennines, but the shape of the swath better resembles that observed compared to RAD. The maximum value close to Livorno, i.e. in the coastal part of the swath, is 158 mm/3h, clearly showing the occurrence of a severe storm.

The LIGHT simulation also shows the local maximum in the southern part of the domain (about 50 mm/3h), but the amount is underestimated.

Figure 17e shows the rainfall forecast by RADLI. The precipitation swath from coastal Tuscany towards NE is more apparent compared to LIGHT and RAD. The maximum rainfall accumulated close to Livorno is 186 mm/3h. Also, the second precipitation maximum in the southern part of the domain reaches 70 mm/3h in good agreement with observations (82 mm/3h). Also, RADLI is the only run producing a satisfactory precipitation field over the south-eastern Emilia Romagna (north-eastern part of the domain), on the lee of the Apennines.

It is also noted that the main precipitation swath forecasted by RADLI is too broad in the direction crossing the swath compared to the observations. This is confirmed by the FBias of RADLI (not shown), which is more than 3 for thresholds larger than 42 mm/3h.

Considering the POD, Figure 17f, we note the considerable improvement given to the score by data assimilation (lightning and/or radar reflectivity factor). POD is larger than 0.5 for RADLI and LIGHT up to the 52 mm/3h thresholds, clearly showing that those two configurations are able to



568 catch the position and timing of the very intense precipitation, especially considering that the
 569 maximum displacement error for the precipitation field is 1.9 km.

570 RAD has a lower capability to correctly forecast the precipitation inland compared to FLA and
 571 RADLI, however: a) it qualitatively reveals the heavy precipitation occurring in the Livorno area; b)
 572 the POD score is considerably improved compared to CTRL.

573 The ETS score, Figure 17g, underlines the good performance of RAD, LIGHT and RADLI compared
 574 to CTRL. RAD has a useful forecast ($ETS > 0$) up to 42 mm/3h, while LIGHT and RADLI show useful
 575 forecast up to 60 mm/3h. The lower ETS of RADLI compared to LIGHT for thresholds larger than 42
 576 mm/3h is caused by the greater number of false alarms occurring in RADLI. The large variations of
 577 the scores for thresholds above 40 mm/3h is caused by the low number of raingauges observing
 578 those rainfall amounts.

579 CTRL has the lowest HR, Figure 17h, up to 16 mm/3h because of the lower number of hits
 580 compared to other configurations. For thresholds larger than 32 mm/3h RADLI has the lowest HR
 581 due to the comparatively higher number of false alarms.

582 In summary, for the most intense precipitation period over Livorno, the data assimilation of
 583 lightning and radar reflectivity factor plays a key role for the correct representation of the storm
 584 intensity, timing and position, giving an improvement of paramount practical importance.

585

586 4.2.3 Livorno: 06-09 UTC 10 September 2017

587 In this period, the most intense phase of the precipitation occurred over Central Italy, over the
 588 coastal part of Lazio (Figure 18a). More in detail, among the 2695 raingauges reporting valid data
 589 over the domain of Figure 18a, 307 reported more than 10 mm/3h, 132 more than 20 mm/3h, 86
 590 more than 30 mm/3h, 66 more than 40 mm/3h, 49 more than 50 mm/3h and 35 more than 60
 591 mm/3h. Among the 35 raingauges measuring more than 60 mm/3h, 33 were over the Lazio
 592 showing the heavy rainfall occurred over the Region.

593 Some precipitation persisted over Tuscany but the rainfall is much lower compared to previous 6h
 594 (the rainfall over Tuscany between 03 and 06 UTC was very intense, not shown). Other notable
 595 precipitation areas are over the NE of Italy (moderate to low amounts), over Central Alps
 596 (moderate values) and over the whole Sardinia (small amounts).

597 Figure 18b shows the rainfall simulated by CTRL. The forecast is unsatisfactory, mainly for the
 598 following two reasons: a) heavy precipitation is simulated over Tuscany (> 75 mm/3h), also close
 599 to the Livorno area; b) very small precipitation is forecast over Central Italy. The rainfall over NE



Italy is well represented in space, but overestimated because the forecast is higher than 50 mm/3h in correspondence of some raingauges, while observed values are 20-25 mm/3h. The small precipitation over Sardinia is not forecasted CTRL.

Considering the evolution of CTRL rainfall forecast for the different phases of the storm, we conclude that CTRL was able to predict abundant rain over Livorno, but this was delayed compared to the real event.

The rainfall simulated by RAD (Figure 18c) clearly improves the forecast compared to CTRL. First, the precipitation over Lazio is very well predicted and the rainfall values are higher than 40 mm/3h (up to 65 mm/3h), so the RAD forecast well represents the main precipitation spot over Italy for this period of time. Second, the precipitation over Tuscany is lowered compared to CTRL, showing the ability of radar reflectivity factor data assimilation to dry the model when it predicts rain that is not observed. Third, the precipitation over Central Alps is represented, even if located about 30 km to the East.

There are also aspects of the rainfall forecast that are less satisfactory: the small precipitation over Sardinia is not represented by RAD; the precipitation over NE Italy is well represented in space but overestimated.

LIGHT forecast, Figure 18d, shows a worse performance compared to RAD for this time period. The precipitation forecast is mainly over Tuscany, where it is overestimated, with a small precipitation spot over Lazio. There are, however, three improvements compared to CTRL and RAD: a) the small precipitation over Sardinia is well represented in LIGHT; b) the precipitation over Central Alps is well predicted; c) the rainfall forecast over NE Italy is overestimated by LIGHT but to a less extent compared to RAD.

The precipitation forecast of RADLI, Figure 18e, represents very well the precipitation over Lazio, and the rainfall amount is better predicted compared to RAD. The precipitation over Sardinia is well represented by RADLI as well as the precipitation over Central Alps, giving the best results among all forecasts.

The POD score (Figure 18f) confirms the above analysis. All the experiments with data assimilation outperform the CTRL forecast, and, for this time period, RAD performs better than LIGHT. RADLI shows the best POD among all configurations because it represents better amount of rainfall over Lazio.

Similar considerations apply to ETS (Figure 18g); it is worth of note the high value of ETS for thresholds larger than 50 mm/3h, which represent heavy rainfall. Again, a forecast that was




missed by CTRL is correctly represented by the assimilation of both radar reflectivity factor and lightning.


The HR score (Figure 18h) shows that CTRL has the lowest score for thresholds below 14 mm/3h because it has a lower number of hits. For higher thresholds (> 32 mm/3h), the impact of the false alarms become important and RADLI has the lowest HR.

637

5. Discussion and Conclusions

In this paper we have shown the impact of the lightning and radar reflectivity factor data assimilation on the very short term forecast (3h) of precipitation for two cases occurred in Italy.

We use  RAMS@ISAC model, whose 3DVar extension to the assimilation of radar reflectivity factor is shown in this paper.

The first case study occurred on 16 September 2017 and it is a moderate case with localised rainfall over  al Italy. It was chosen because the control forecast, i.e. without radar reflectivity factor or lightning data assimilation, missed the event. The second event, occurred on 9-10 September 2017, was characterised by exceptional rainfall over several parts of Italy. This event was partially represented by the control forecast. In particular, the forecast of the event was incorrect because: a) the control forecast was delayed compared to the observations; b) the control forecast missed the rainfall over central Italy (Lazio Region).

It is important to recall that the impact of the lightning data assimilation on the precipitation forecast of RAMS@ISAC was already studied for the HyMeX-SOP1 period (Federico et al., 2017a, 2017b), and a robust statistic is already available. The results of this study confirm the important role of the lightning data assimilation on the rainfall forecast for two case studies. However, considering the assimilation of radar reflectivity factor, and its combination with lightning data assimilation in RAMS@ISAC, the results of this paper are new.

Because we analysed only two case studies, no definitive conclusions can be derived on the performance of RAMS@ISAC for radar reflectivity factor data assimilation. There are, however, few points worth of mention.

The VSF performance of RAMS@ISAC is systematically improved by the assimilation of radar reflectivity factor. This improvement is of paramount importance for some specific VSF (for example for the 00-03 UTC for Livorno), when the control forecast missed the event while it was correctly predicted by radar reflectivity factor data assimilation. Sometimes the improvement of reflectivity factor data assimilation has a lower impact on the precipitation forecast, as for the



664 period 18-21 UTC on 9 September 2017 (Livorno) or for the second stage (18-21 UTC) of the
 665 Serano case; however, also for these cases the assimilation of reflectivity improves the
 666 precipitation forecast.

667 Lightning and radar observations are different and both add value to the VSF. In particular, flashes
 668 are recorded when deep convection develops, while radar reflectivity factor is observed also for
 669 light stratiform rain. Flashes are available for the open sea, while radar reflectivity factor is
 670 confined to the range of the coastal radars in the network. Lightning have a seasonal dependence
 671 over Italy, with the maximum in summer and fall, while radar reflectivity factor is available in all
 672 seasons.

673 For the above reasons, the impact of the two kinds of data on the rainfall VSF is expected
 674 different. Some examples have been shown: the light precipitation over Northern Italy for Serano
 675 case is well forecasted assimilating radar reflectivity factor, while it is not simulated assimilating
 676 flashes because they are too few in this area to force convection into the model; lightning data
 677 assimilation is able to better represent the deep convection occurring during the intense phase of
 678 the Livorno case (00-03 UTC), especially because it is able to force convection where it occurs,
 679 reducing false alarms. The last characteristic has been found in some others VSF of the case
 680 studies considered, and it is shown by the fact that the ETS score for LIGHT is sometimes the best
 681 among all simulations.

682 The model configuration assimilating both radar reflectivity factor and lightning (RADLI) is able to
 683 retain important features of both data assimilation. For example, the simulation of the Livorno
 684 case in the phase 06-09 UTC was able to simulate the heavy precipitation over Lazio thanks to the
 685 radar reflectivity factor data assimilation and the precipitation over Sardinia, as well as the
 686 moderate precipitation over Central Alps, thanks to the lightning data assimilation.

687 Another example of synergistic interaction between the two types of data assimilation was found
 688 for the most intense phase of the Serano case (03-06 UTC on 16 September 2017). In this period,
 689 the light precipitation over the Alps was forecasted by RADLI because of the assimilation of radar
 690 reflectivity factor, while the localised precipitation maximum over Central Italy was better forecasted
 691 thanks to the synergistic action of lightning and reflectivity factor data assimilation.

692 The property of RADLI to retain the precipitation forecast features of both data is shown by the
 693 POD score, which is the best, for most cases and thresholds, for RADLI.



694 Another interesting feature is the considerable improvement of the POD of RADLI compared to
695 CTRL for the lowest threshold, showing the better ability of RADLI to predict the area where
696 precipitation will occur at the short term.

697 It is also underlined that the data assimilated, both lightning and radar reflectivity factor, are
698 produced operationally and available in real time and could be used for an operational
699 implementation of the model.

700 All the above features are promising and deserve future studies to better understand the role of
701 radar reflectivity factor and its interaction with lightning data assimilation to improve the
702 precipitation forecast; there are, however, less satisfactory aspects of assimilating both radar
703 reflectivity factor and lightning data. The RADLI forecast has more false alarms compared to RAD
704 and LIGHT and this penalises the usefulness of RADLI forecast. This is shown by the lower ETS and
705 HR score of RADLI, especially compared to LIGHT, for some thresholds and VSF, despite the larger
706 values of the POD of RADLI.

707 The RADLI forecast can miss intense precipitation: this is shown, for example, by the VSF of 18-21
708 UTC on 9 September 2017 (Livorno), when RADLI underestimated the most intense phase of the
709 storm in Livorno.

710 In addition to the acquisition of more case studies, there are two directions of future development
711 of this work. The lightning data assimilation can be formulated by 3DVar, using a strategy similar
712 to the radar reflectivity factor in which pseudo-profiles of relative humidity are first generated
713 where flashes are recorded, and then those profiles are assimilated by 3DVar. This methodology
714 was already reported in Fierro et al. (2016). The assimilation of both radar reflectivity factor and
715 lightning using 3DVar will be explored in future studies.

716 Another important point to study is how long the innovations introduced by data assimilation lasts
717 in the model forecast. While in this study we explored the VSF at 3h, future studies must explore
718 longer time ranges. A similar study was performed for lightning data assimilation (Federico et al.,
719 2017b), using a model set-up very similar to that used in this paper. Results showed that the
720 lightning data assimilation gave a small and positive contribution to the precipitation forecast up
721 to 24 h. However, the impact of data assimilation decreased rapidly, and the improvement of the
722 rainfall forecast was significant after 6h, small after 12h and negligible after 24 h. A study
723 considering both radar reflectivity factor and lightning should be performed to understand the
724 resilience of the innovations introduced by data assimilation.

725



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895 TABLES

896 Table 1: List of physical parameterisations used for RAMS@ISAC in this paper.

Physical parameterization	Selected scheme
Parametrized cumulus convection	Modified Kuo scheme to account for updraft and downdraft (Molinari and Corsetti, 1985). The scheme is applied to R10 only.
Explicit precipitation parameterization	Bulk microphysics with six hydrometeors (cloud, rain, graupel, snow, ice, water vapor). Described in Hong and Lim (2006).
Exchange between the surface, the biosphere and atmosphere.	LEAF3 (Walko et al., 2000). LEAF includes prognostic equations for soil temperature and moisture for multiple layers, vegetation temperature and surface water, and temperature and water vapor mixing ratio of canopy air.
Sub-grid mixing	The turbulent mixing in the horizontal directions is parameterised following Smagorinsky (1963), vertical diffusion is parameterised according to the Mellor and Yamada (1982) scheme, which employs a prognostic turbulent kinetic energy.
Radiation scheme	Chen-Cotton (Chen and Cotton, 1983). The scheme accounts for condensate in the atmosphere.

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919 Table 2: Basic parameters of the RAMS@ISAC grids (R10, R4 and R1, corresponding, respectively, to the domains D1,
 920 D2 and D3). NNXP is the number of grid points in the WE direction, NNYP is the number of grid-points in the NS
 921 direction, NNZP is the number of vertical levels, DX is the size of the grid spacing in the WE direction, DY is the grid
 922 spacing in the SN direction. Lx, Ly, and Lz are the domain extensions in the NS, WE, and vertical directions. CENTLON
 923 and CENTLAT are the coordinates of the grid centres.

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	R10, D1	R4, D2	R1, D3
NNXP	301	401	203
NNYP	301	401	203
NNZP	36	36	36
Lx	3000 km	1600 km	~270 km
Ly	3000 km	1600 km	~270 km
Lz	~22400 m	~22400 m	~22400 m
DX	10 km	4 km	4/3 km
DY	10 km	4 km	4/3 km
CENTLAT (°)	43.0 N	43.0 N	43.7 N
CENTLON (°)	12.5 E	12.5 E	11.0 E



FIGURES

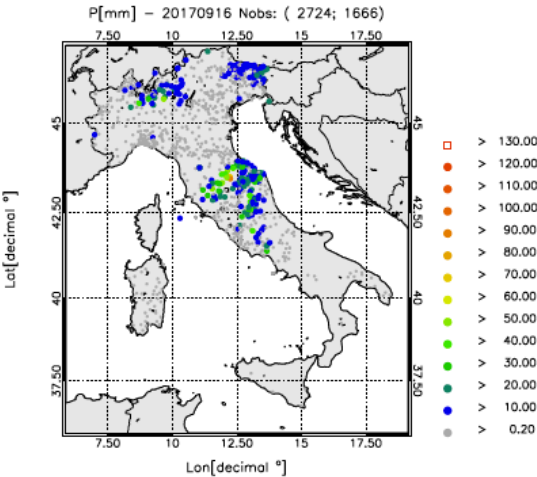
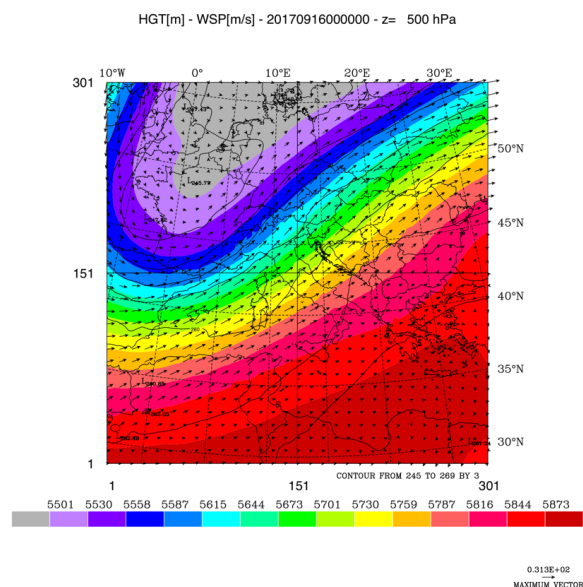


Figure 1: Daily precipitation (P) [mm] over Italy from the network of the Department of Civil Protection on 16 September 2017. Only raingauges observing at least 0.2 mm/day are shown. The first number in the figure title within brackets represents the available raingauges, while the second number represents raingauges observing at least 0.2 mm/3h. The lowest precipitation class is represented by smaller dots, the largest by a red square.



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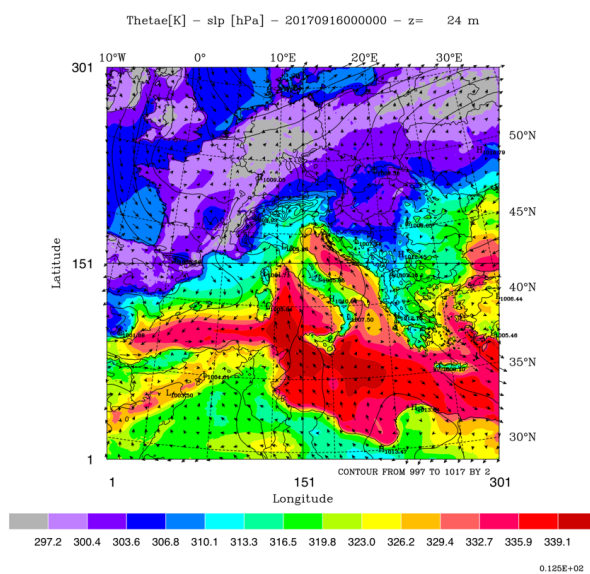
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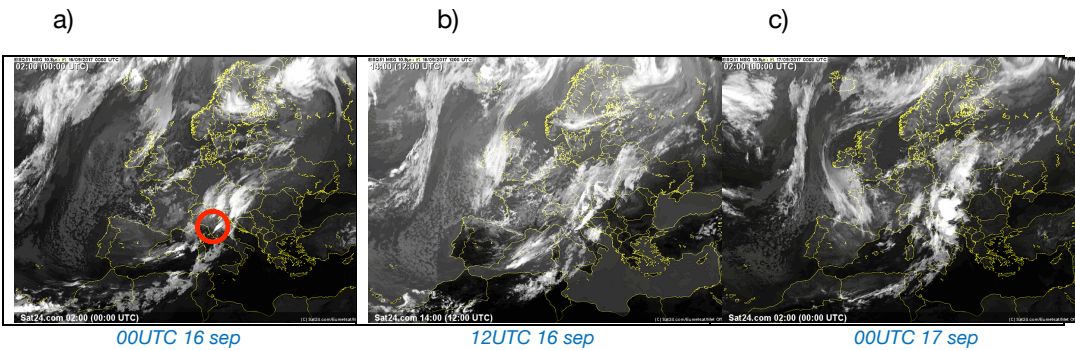


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983 Figure 2: a) Geopotential height (filled contours), temperature (contours) and wind vectors at 500 hPa at 00 UTC on 16
 984 September 2017. Maximum velocity is 31 m/s; b) equivalent potential temperature (filled contours), sea-level
 985 pressure (contours) and wind vectors at 24 m above the surface (first vertical level, maximum value 13 m/s). A low-
 986 pressure pattern is forming over northern Italy, with a front in the western Mediterranean.

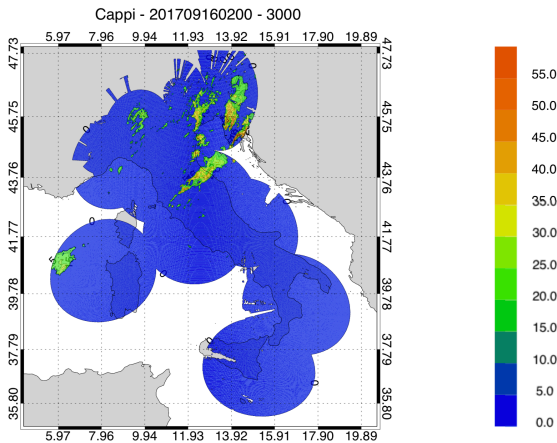


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Figure 3: a) Satellite images (METEOSAT second generation) of the infrared channel, 10.8 micron, for 00 UTC and 12 UTC on 16 September, and for 00 UTC on 17 September 2017. A well-defined cloud system is apparent inside the red circle of the image at 00 UTC on 16 September 2017.



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Figure 4: National radar mosaic at 3 km above the sea level observed at 02 UTC on 16 September 2017.

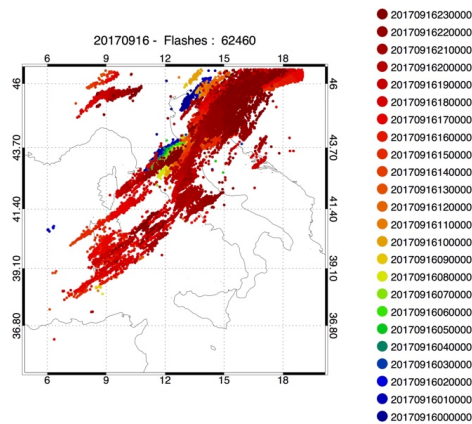


Figure 5: Lightning recorded on 16 September 2017. The total number of flashes recorded is shown in the title. Different colours represent the time (UTC) of occurrence of the lightning.

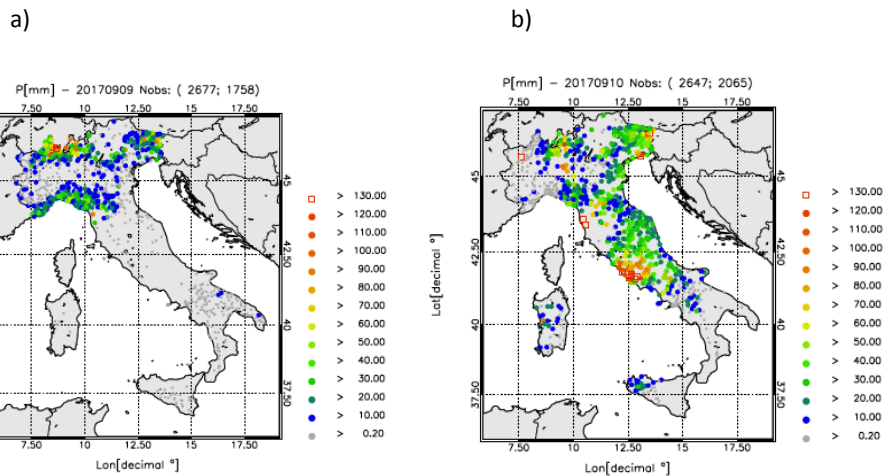
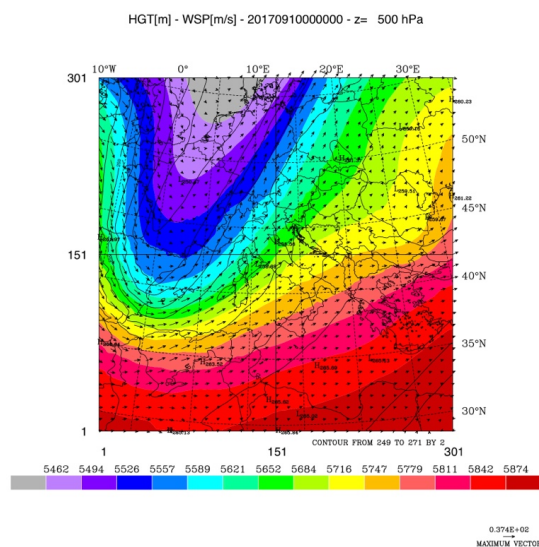


Figure 6: a) As in Figure 1 but for a) 9 September 2017 and b) 10 September 2017.



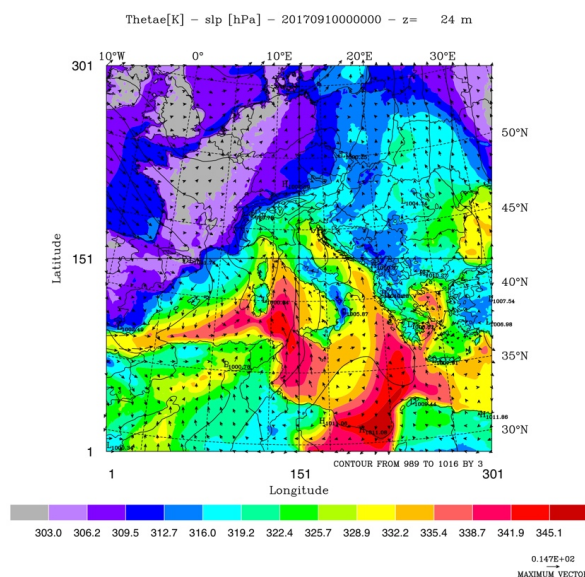
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1014 a)



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1018 Figure 7: a) Geopotential height (filled contours), temperature (contours) and wind vectors at 500 hPa at 00 UTC on 10
 1019 September 2017. Maximum velocity is 37 m/s; b) equivalent potential temperature (filled contours), sea-level
 1020 pressure (contours) and wind vectors at 24 m above the surface (first vertical level, maximum value 15 m/s).
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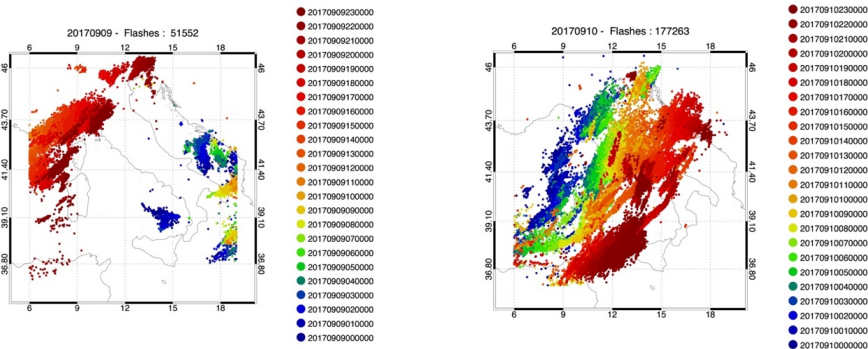


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1027 Figure 8: a) Lightning recorded on 09 September 2017; b) Lightning recorded on 10 September 2017. The number of
1028 flashes recorded on each day is shown in the title. Different colours represent the time of occurrence of the lightning.

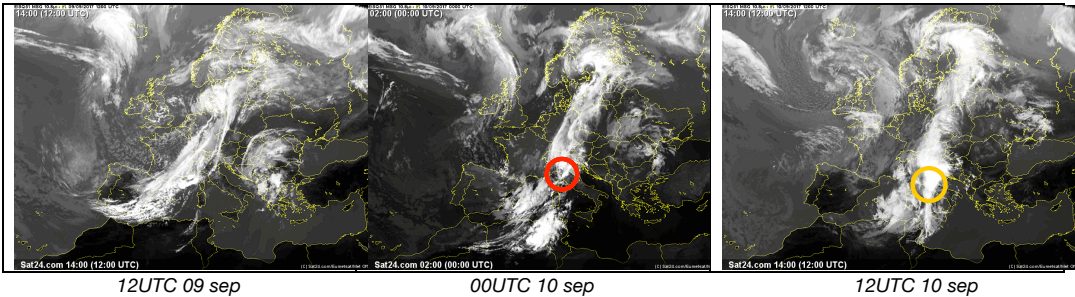
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1032 Figure 9: a) Satellite images (METEOSAT second generation) of the infrared channel, 10 micron, for 12 UTC on 9
1033 September 2017, 00 UTC and 12 UTC on 10 September 2017.

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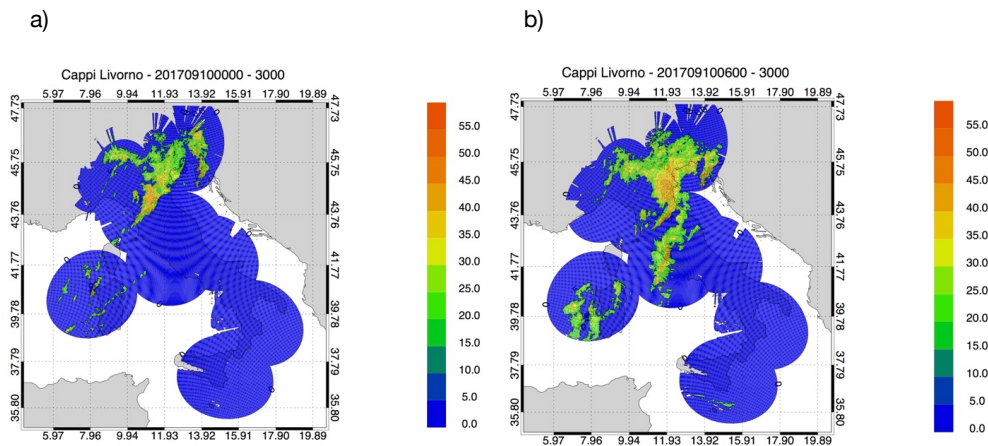
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Figure 10: a) National radar mosaic at 3 km above the sea level observed at 00 UTC on 10 September 2017; b) as in a) for the 06 UTC.

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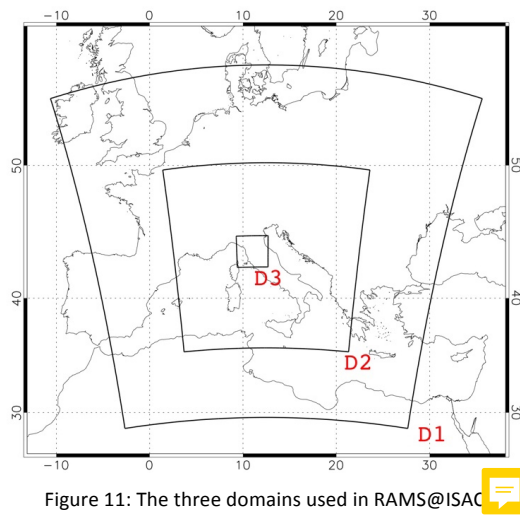
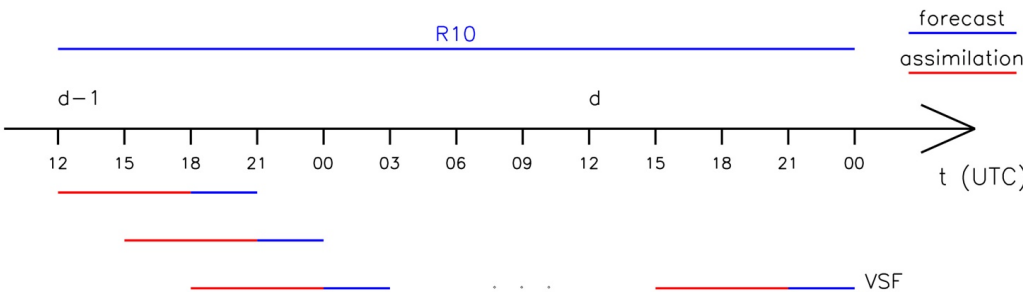


Figure 11: The three domains used in RAMS@ISAC



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Figure 12: The time implementation of the RAMS@ISAC very short-term forecast.



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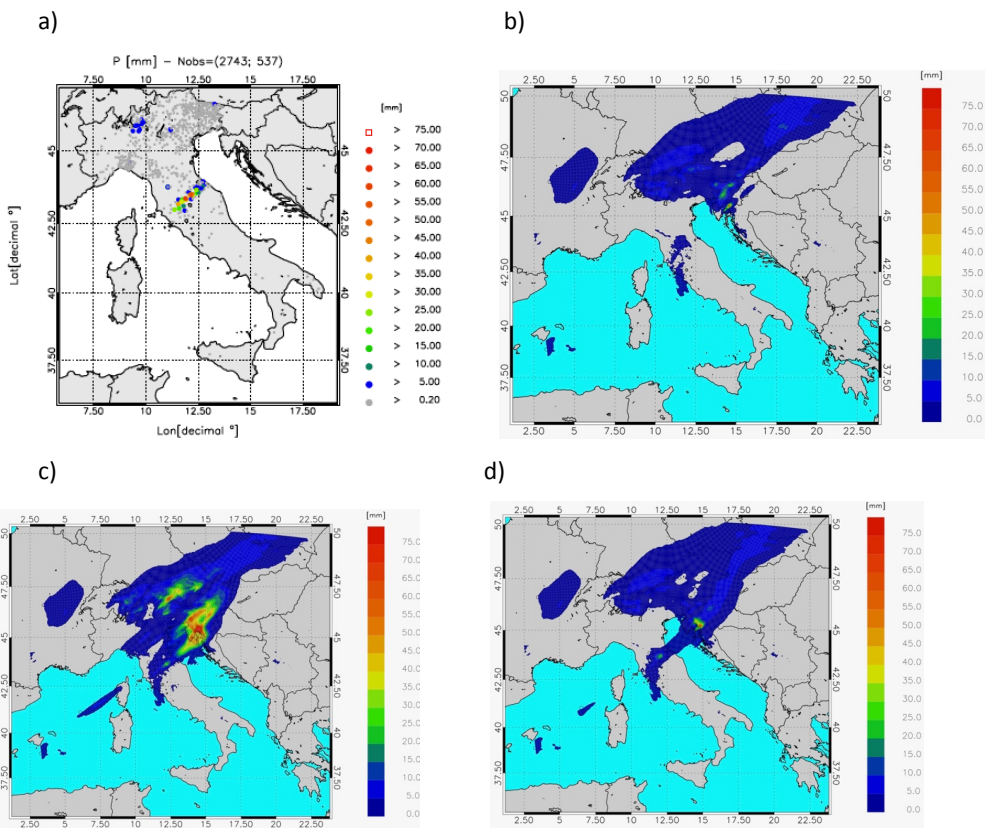


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1057 Figure 13: The radar network of the Department of Civil Protection. Green radars operate with dual-polarisation, blue
1058 radars have single polarisation.

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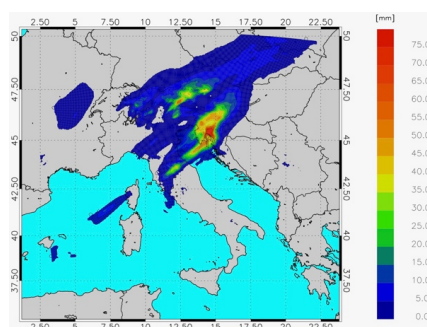
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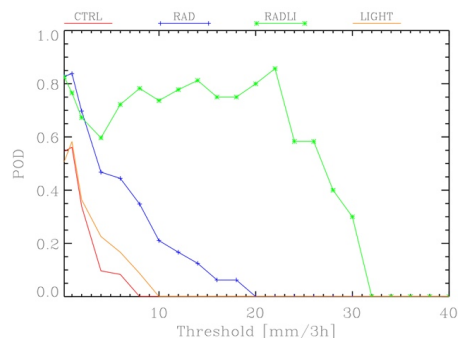
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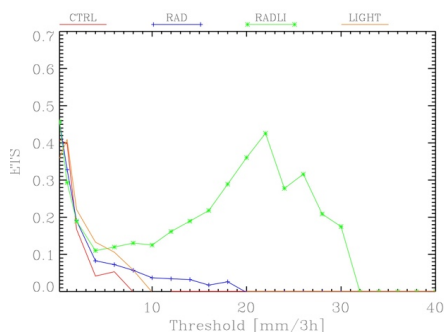
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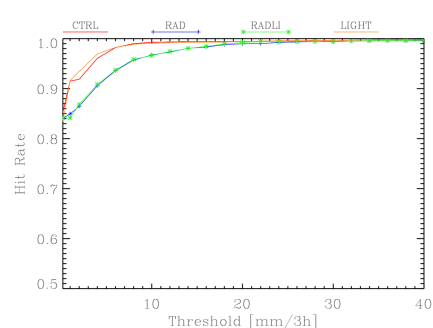


Figure 14: a) rainfall reported by raingauges between 03 and 06 UTC on 16 September 2017. Only raingauges observing at least 0.2 mm/day are shown. The first number in the title within brackets represents the available raingauges, while the second number represents those observing at least 0.2 mm/3h; b) as in a) for the CTRL forecast; c) as in a) for the RAD forecast; d) as in a) for the LIGHT forecast; e) as in a) for the RADLI forecast; f) POD score for the period 03-06 UTC on 16 September 2017; g) as in f) for the ETS score. POD and ETS scores are computed over the domain of Figure 14a

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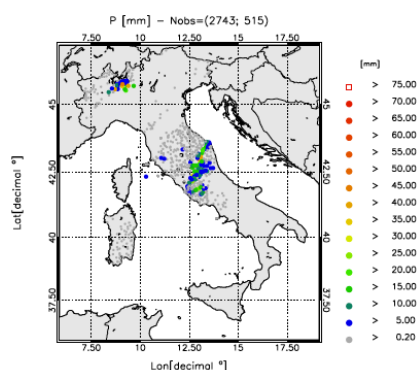
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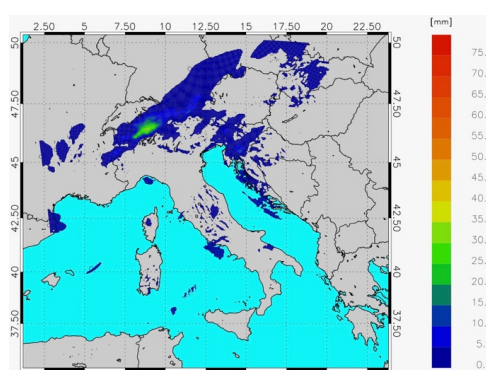


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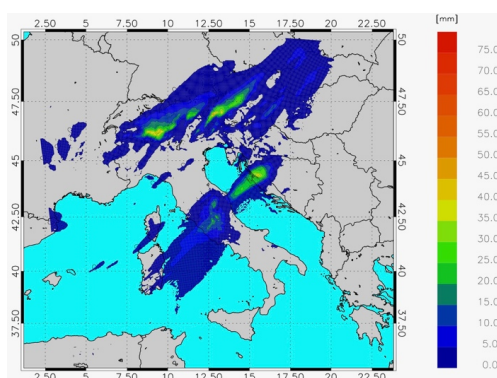


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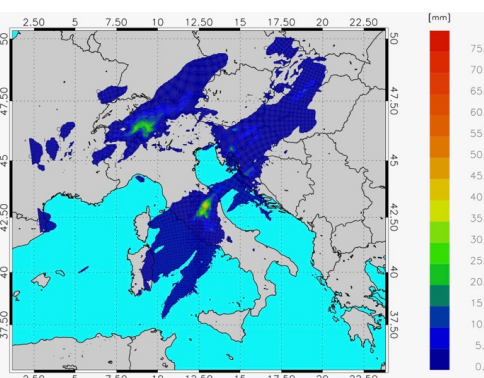


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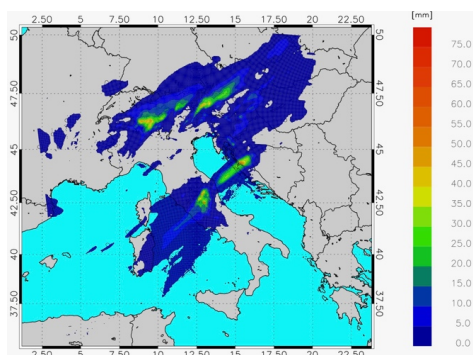


d)

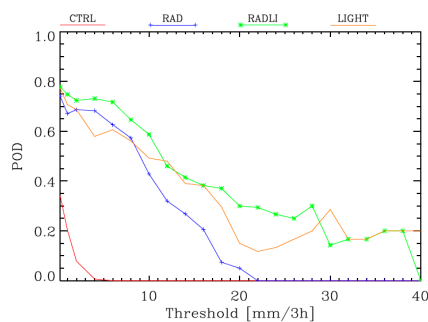


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e)



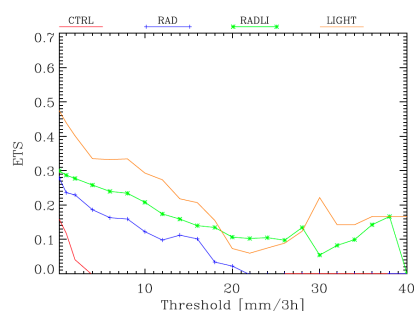
f)



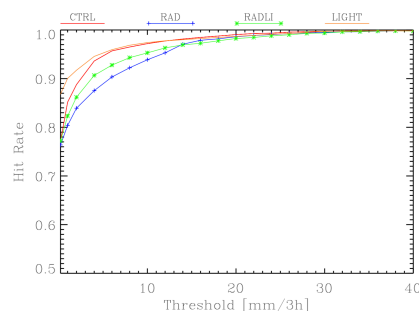
1093
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1099 g)



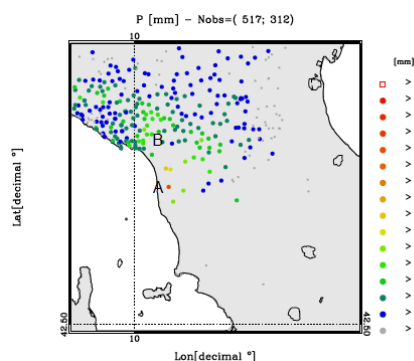
h)



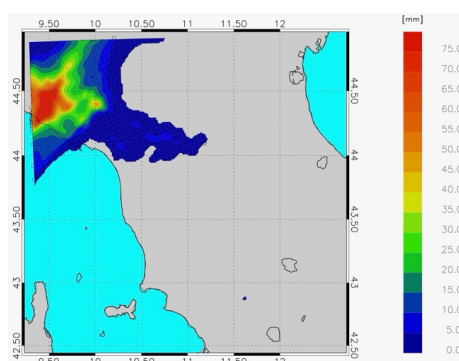
1100

1101 Figure 15: a) rainfall reported by raingauges between 18 and 21 UTC on 16 September 2017. Only raingauges
 1102 measuring at least 0.2 mm/3h are shown. The first number in the title within brackets represents the available
 1103 raingauges, while the second number represents those observing at least 0.2 mm/3h; b) as in a) for the CTRL forecast;
 1104 c) as in a) for the RAD forecast; d) as in a) for the LIGHT forecast; e) as in a) for the RADLI forecast; f) POD score for the
 1105 period 18-21 UTC on 16 September 2017; g) as in f) for the ETS score. POD and ETS scores are computed over the
 1106 domain of Figure 15a.

1107 a)

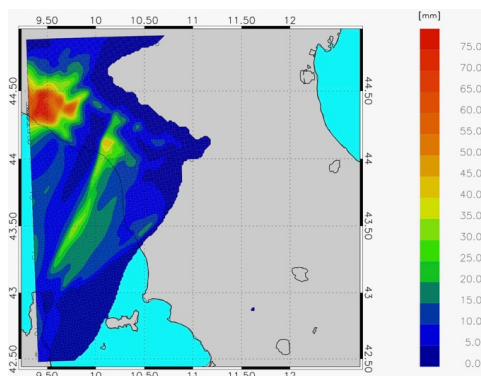


b)

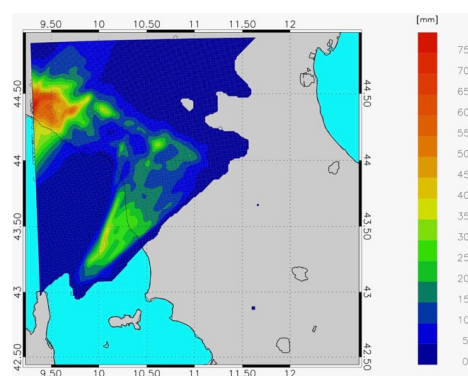


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1111 c)



d)



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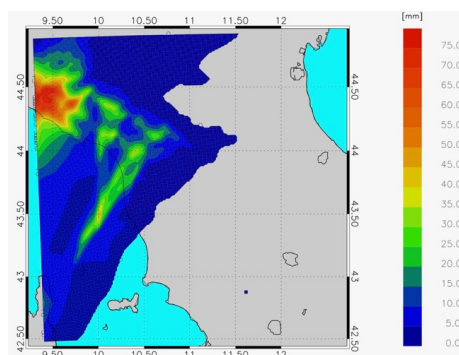
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1115

e)

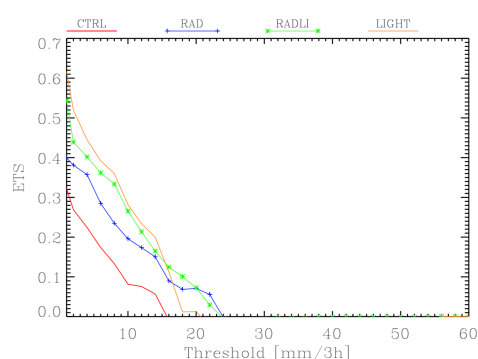


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g)



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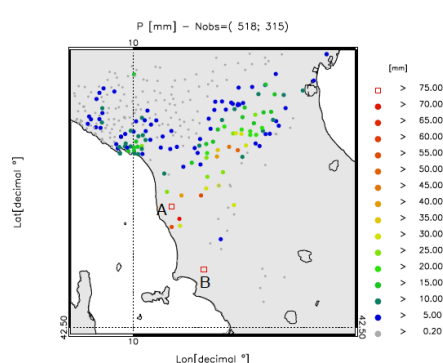
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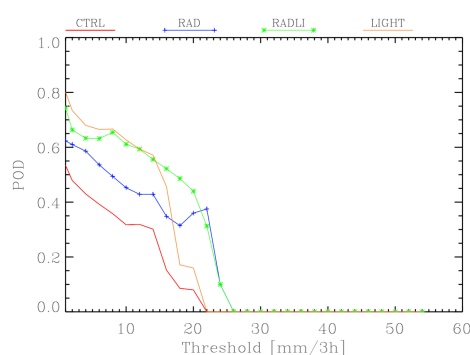
a)



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1129

f)



h)

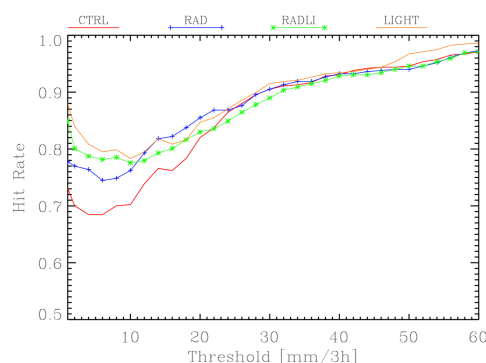
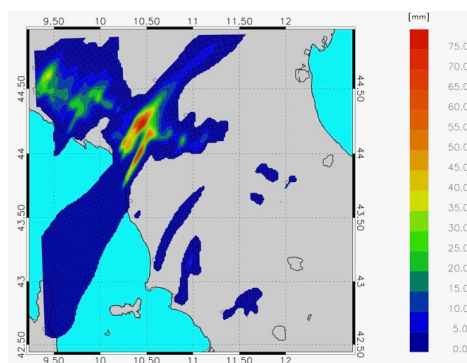


Figure 16: a) rainfall reported by raingauges between 18 and 21 UTC on 9 September 2017. Only raingauges observing at least 0.2 mm/3h are shown. The first number in the title within brackets represents the available raingauges, while the second number represents those observing at least 0.2 mm/3h; b) as in a) for the CTRL forecast; c) as in a) for the RAD forecast; d) as in a) for the LIGHT forecast; e) as in a) for the RADLI forecast; f) POD score for the period 18-21 UTC on 9 September 2017; g) as in f) for the ETS score; h) as in f) for the Hit Rate. Panel A in Figure 16a shown the position of Livorno and the domain of Figure 16a.

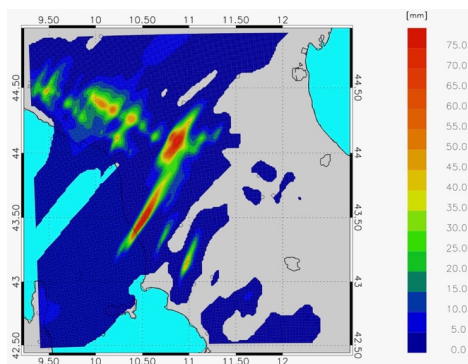
b)





1130

c)



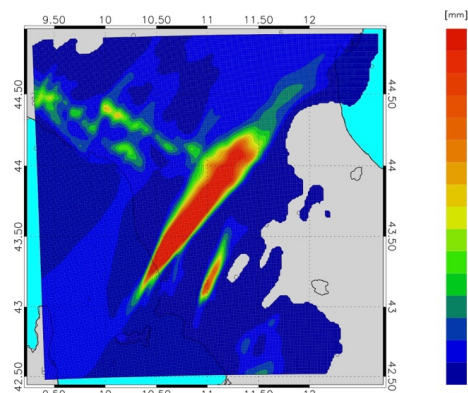
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e)

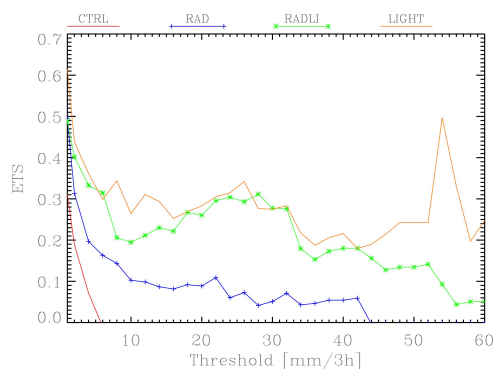


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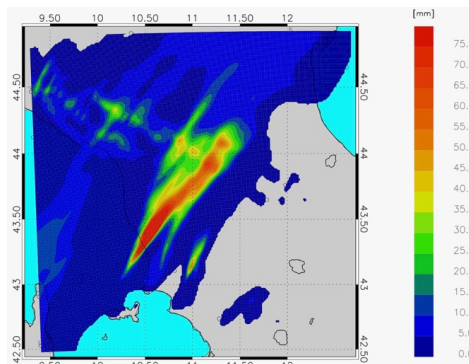
1137 g)

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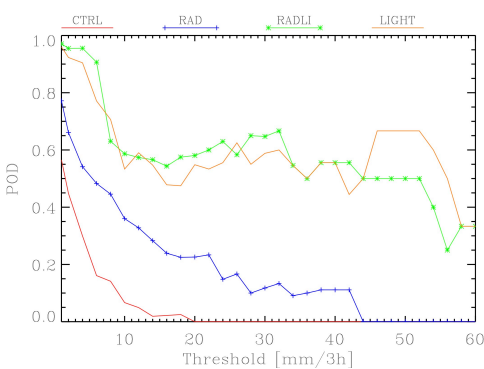


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d)



f)



h)

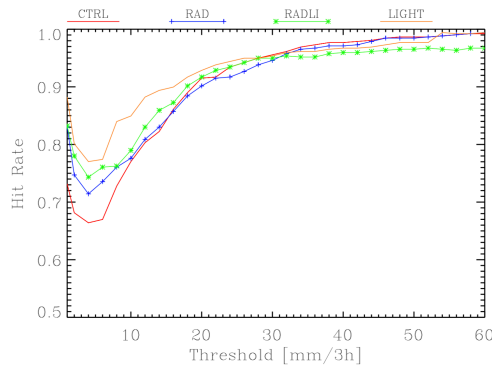
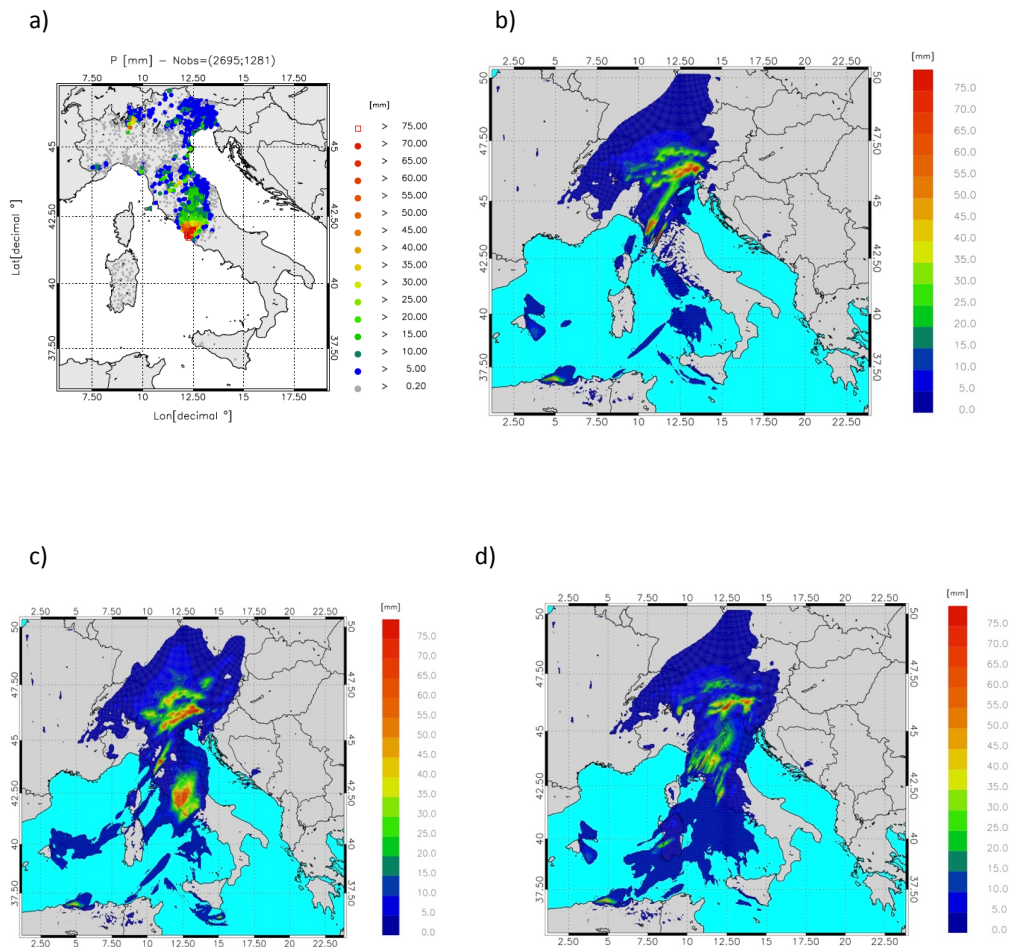




Figure 17: a) rainfall reported by raingauges between 00 and 03 UTC on 10 September 2017. Only stations reporting at least 0.2 mm/3h are shown. The first number in the title within brackets represents the number of raingauges available over the domain, while the second number shows those observing at least 0.2 mm/3h; b) as in a) for the CTRL forecast; c) as in a) for the RAD forecast; d) as in a) for the LIGHT forecast; e) as in a) for the RADLI forecast; f) POD score for the period 00-03 UTC on 10 September; g) as in f) for the ETS score. POD and ETS scores are computed over the domain of Figure 17a.



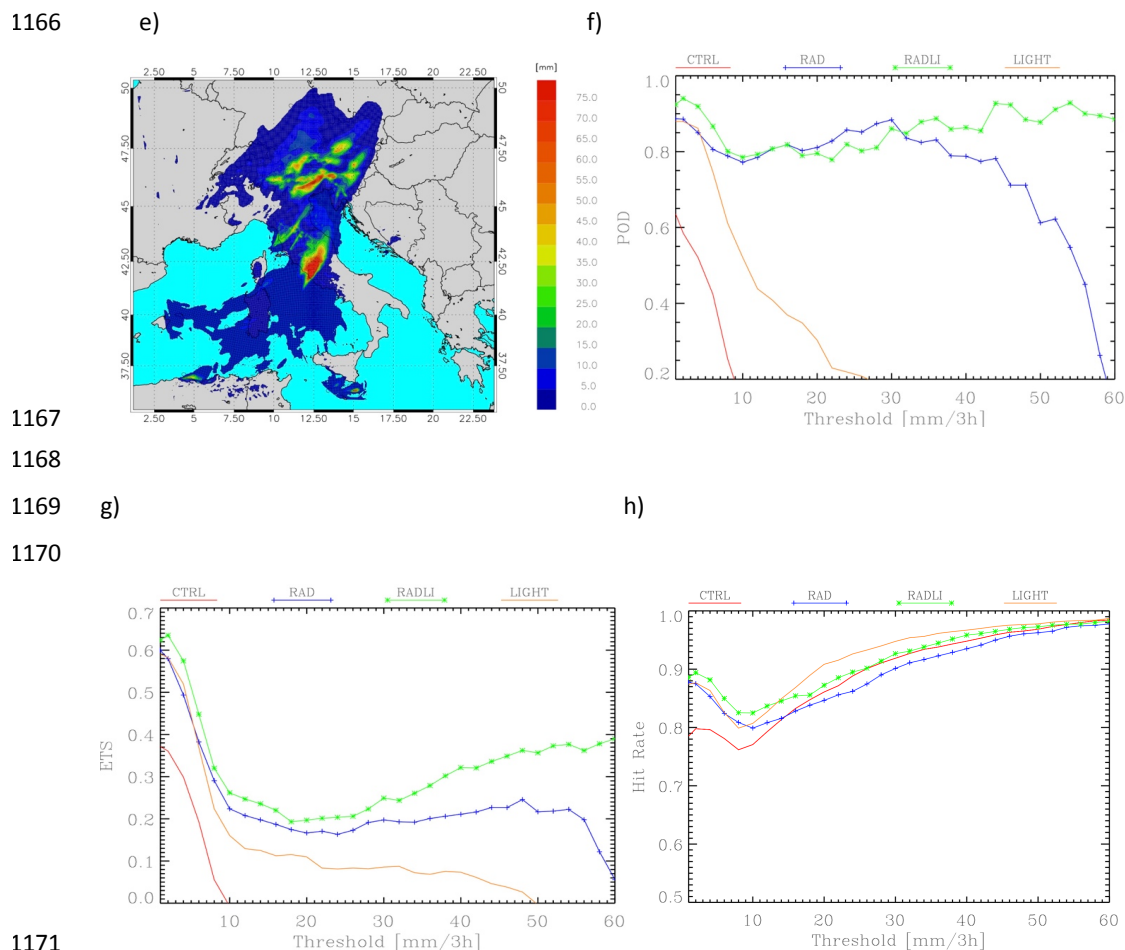


Figure 18: a) rainfall reported by raingauges between 06 - 09 UTC on 10 September 2017. For this time period 2695 raingauges reported valid observations in the domain, however only stations reporting at least 0.2 mm/3h are shown. The first number in the title within brackets represents the number of raingauges available over the domain, while the second number shows those observing at least 0.2 mm/3h; b) as in a) for the CTRL forecast; c) as in a) for the RAD forecast; d) as in a) for the LIGHT forecast; e) as in a) for the RADLI forecast; f) POD score for the period 06-09 UTC on 10 September; g) as in f) for the ETS score; h) as in f) for the Hit Rate. POD and ETS scores are computed over the domain of Figure 18a.