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High-resolution prediction of a major convective period over West Africa

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In this article, we evaluate the predictions of the French cloud-resolving model AROME using a set of high-resolution (5 km) simulations that focus on the well documented African Monsoon Multidisciplinary Analysis (AMMA) period of 23–28 July 2006 over a large domain $(0-22^{\circ}N, 15^{\circ}W-20^{\circ}E)$. The model skill is assessed against independent Global Positioning System observations of precipitable water and in terms of quantitative precipitation forecasts. As the rain-gauge network is sparse over West Africa, the simulated precipitation fields were compared with data from satellite-based precipitation products (TRMM-3B42). We show that initial and boundary conditions significantly improve the AROME forecasts when the large-scale forcing model (ARPEGE) assimilates surface-sensitive observations from microwave remote-sensing sensors over land surface.

The daily mean AROME precipitation shows a spatial distribution in good agreement with the satellite precipitation estimates. The intertropical convergence zone is correctly reproduced in terms of shape and location but its intensity is broadly overestimated by about 25%. The AROME model is shown to be able to reproduce all regimes, from light rain to the biggest Mesoscale Convective Systems (MCSs). The observations made at the Niamey AMMA supersite allow a detailed evaluation. Near the Niamey AMMA supersite, we show that AROME is able to represent most of the key features of the West African monsoon from the diurnal to synoptic scales. The life cycle of two successive sequences of MCSs associated with an African easterly wave and a deep monsoon burst are well captured by AROME.

Finally, a tracking approach based on the 1 h accumulated precipitation is applied to both Global Satellite Mapping of Precipitation (GSMaP) satellite estimates and to AROME and ARPEGE forecasts, allowing a good characterization of each MCS and statistics. Contrary to ARPEGE, the AROME MCSs trajectories and lifetimes, and the diurnal cycles of their initiation and dissipation, are in agreement with the GSMaP tracking and previous MCS statistics.

Key Words: cloud-resolving simulation; mesoscale convective system; West African monsoon; African easterly wave; tracking; prediction

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51 1. Introduction

West African populations are extremely dependent on rainfall 53 provided by the short monsoon period for their agriculture, 54 health and economy. The region most sensitive to seasonal 55 rainfall is undoubtedly the Sahel band extending between 12° 56 and 18°N, on the northern flank of the intertropical conver-57 gence zone (ITCZ) and on the southern rim of the Sahara 58 Desert. This region is characterized by a high variability at 59 decadal, interannual and intraseasonal scales (Janicot et al., 2011; 60 Rodriguez-Fonseca et al., 2011). Understanding and modelling 61

the West African rainfall variability (West African monsoon, WAM) was one of the major aims of the African Monsoon Multidisciplinary Analysis (AMMA: http://www.ammainternational.org/spip.php?rubrique1) project (Redelsperger *et al.*, 2006).

Mesoscale convective systems (MCSs) are responsible for much of the rainfall over West Africa (Omotosho, 1985; Mathon and Laurent, 2001), providing 80% of the rainfall over the Sahel (Mathon *et al.*, 2002). However, weather and climate models with typical horizontal resolutions between several tens and hundreds of kilometres only partly resolve the largest MCSs and have poor

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skill in forecasting the amount and variability of their associated precipitation (Lebel et al., 2000; Roehrig et al., 2013). African easterly waves (AEWs) are the major synoptic weather systems in the WAM, developing in the vicinity of the midlevel African Easterly Jet over the (Reed *et al.*, 1977; Thorncroft and Hoskins, 1994a, 1994b, and significantly modulating rainfall.

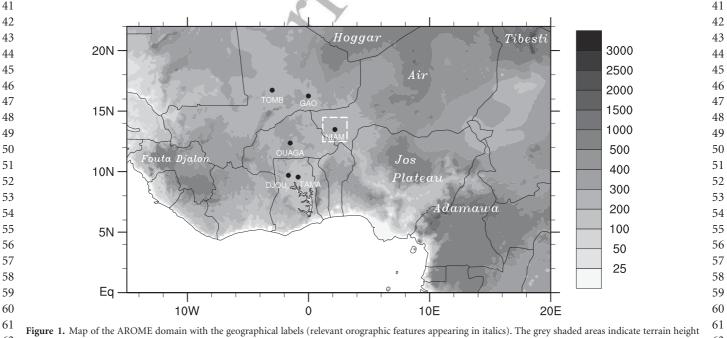
6 In comparison with midlatitudes, the improvement in 7 numerical weather prediction (NWP) in the Tropics has been 8 very slow even when metrics better suited to low latitudes, 9 characterized by a large Rossby radius (Berry et al., 2006), are used. 10 The skill of an AEW forecast within different operational NWPs is less than 48 h (Torn, 2010), and the synoptic variability and 11 the coupling between AEWs and precipitation differ considerably 12 among NWP systems. The NWP forecasts are usually unable to 13 generate new AEWs, tending to advect those present in the initial 14 conditions. The basic reasons for such biases and poor skills at 15 short forecast ranges are multiple and depend on the quality of 16 observations (Thorncroft et al., 2003) and the data assimilation 17 (Tompkins et al., 2005; Agusti-Panadera et al., 2010), as well as the 18 models, but also on the level of predictability in West Africa. From this perspective, mesoscale models may be more appropriate than 19 large-scale models for studying and forecasting AEWs, MCSs and 20 their complex interplay. 21

In the 1990s, technological advances allowed cloud-resolving 22 models (CRMs) to become a powerful tool for the study of 23 MCSs and associated AEWs, especially in West Africa (Diongue 24 et al., 2002). The application of CRMs has been confined to 25 research studies on processes until recently, because of computer 26 power limitations. During the last decade, a new generation of 27 convection-permitting operational models have been developed (WRF in the USA, MetUM LAM in the United Kingdom, AROME 28 in France, and COSMO in Germany, among many others) and 29 are now able to properly resolve the life cycles of MCSs. The 30 initial and boundary conditions appear to significantly control 31 the location of the rainfall over West Africa at the synoptic 32 scale for 1-2 day forecasts (Guichard et al., 2010; Torn, 2010). 33 Some sensitivity studies suggest that the explicit representation of 34 deep convection improves the forecast of AEWs and significantly 35 improves the predicted rainfall distribution, trajectories, and 36 propagation speed of MCSs. Up to now, only a few observed 37 cases of MCS have been simulated with CRM over a period of 38 several days in West Africa (Barthe et al., 2010; Birch et al., 2012; Laing et al., 2012). In the framework of the Cascade UK project, 39 simulations were carried out with a nesting approach (described 40

in Lean et al. (2008)) for 10-day periods (Pearson et al., 2010; 1 Marsham *et al.*, 2011) over large domains (several thousand km) 2 to better understand the interactions between tropical convection 3 and the larger scales.

4 The skill of such CRM simulations depends strongly on the 5 quality of the large-scale atmospheric and surface fields used to 6 initialize and force simulations. It is still a challenge to attain 7 such accurate simulations over Africa due to the scarcity of 8 observations, the large Rossby radius and the major role played 9 by diabatic and nonlinear processes such as convection. A major legacy of AMMA has been the update and reinforcement of 10 the West African operational network of radiosondes (Parker 11 et al., 2008) completed by a wide range of surface, remote-12 sensing and aircraft observations. It provides a unique set of data 13 documenting the WAM system at different scales. The AMMA 14 programme allowed substantial progress to be made in remote-15 sensing data use and assimilation in tropical regions (Bock et al., 16 2007; Bauer, 2009; Faccani et al., 2009; Karbou et al., 2010b). It is 17 worth mentioning that a network of six Global Positioning System (GPS) ground stations was developed over West Africa (Figure 1) 18 during the special observing period (SOP). These observations 19 were found very useful for measuring the total column water 20 vapour (TCWV) at different scales (Bock et al., 2007) and for 21 evaluating NWP analyses and forecasts. 22

The present study focuses on the well-documented 23-28 23 July convective period during the AMMA 2006 field experiment 24 (Barthe et al., 2010). The large AMMA-SOP dataset at the Niamey 25 AMMA supersite, in particular the ultra-high-frequency (UHF) 26 and Massachusetts Institute of Technology (MIT) Doppler radar 27 and the dense network of rain-gauges, was used to evaluate the fine-scale NWP Applications of Research to Operations 28 at MEsoscale (AROME) model (Seity et al., 2011). Although 29 AROME has been in routine use over western Europe for 30 four years, this was the first time it had been used over a broad 31 West African domain during the monsoon to explore its forecast 32 skills. The main objective of this article is to evaluate the AROME 33 model over West Africa during the WAM season and to assess 34 the impact of some improvements of the initial and coupling 35 atmospheric fields resulting from the assimilation of remotesensing microwave observations (Karbou et al., 2010a, 2010b). 36 The analyses and forecasts were also evaluated against the GPS 37 observations and Tropical Rainfall Measuring Mission (TRMM-38 3B42) rainfall estimates (Huffman et al., 2007). In addition to 39 the scores, a precipitation-features tracking algorithm based on 40



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62 62 above mean sea level (m). The dark circles represent the six GPS stations (TOMB for Timbuktu (3°W, 16.73°N), Gao (0°, 16.25°N), NIAM for Niamey (2.18°E, 13.48°N), OUAGA for Ouagadougou (1.51°W, 12.36°N), DJOU for Djougou (1.66°E, 9.69°N), TAMA for Tamale (0.86°W, 9.55°N). The white dotted box represents 63 63 the boundaries of the domain called 'Niamev box' used in Figure 13. 64

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1 1 h accumulated precipitation is used over the period of interest to evaluate the forecast skill and the differences of behaviour 2 of individual models. This tracking approach provides objective 3 statistics on the life cycles of the MCSs, such as their diurnal 4 initiation and dissipation and their life span. 5

Section 2 presents the case-study, the AROME model, the 6 design of the experiments, the data and the evaluation procedure. 7 In section 3, the AROME sensitivity to initial and coupling 8 atmospheric fields from the ARPEGE model is discussed, 9 depending on the ARPEGE configuration and on the assimilation 10 of Advanced Microwave Sounding Unit (AMSU-B) microwave channels. In section 4, we evaluate the ability of the AROME 11 model to represent the life cycles of MCSs. Conclusions are 12 drawn in section 5, and we discuss the perspectives provided by 13 such CRM simulations for studying the life cycle of MCSs and 14 their interplay with AEWs. 15

16 2. The case-study and experimental design 17

18 2.1. Case-study 19

20 To evaluate the AROME simulations, we looked at a sequence 21 of convective events that occurred near Niamey during the 2.2 period of 25-26 July 2006. It was a period of particularly active 23 convection embedded in a large-scale circulation favourable to convection and characterized by a monsoon surge combined 24 with the passage of an AEW. Barthe et al. (2010) studied the 25 environment of two successive mesoscale convective systems 2.6 (MCS1 and MCS2 hereafter) using the large AMMA-SOP dataset 27 (UHF radar, MIT Doppler radar, rain-gauges, GPS stations, 28 etc.). The long-lived MCSs were located ahead of, and in 29 phase with, the trough of the AEW, in phase with a strong 30 vertical shear and the intense monsoon surge penetrating up to 31 18°N (Cuesta et al., 2010). From the afternoon of 26 July and for 36 h, convection developed with suppression of organized 32 convection and the occurrence of isolated convective cells. The 33 local convection resulted from some favourable factors (humidity, 34 convective available potential energy (CAPE), weak convective 35 inhibition (CIN)) that triggered convection, while inhibiting 36 factors (midlevel dry layer, weaker low-level wind shear pointing 37 to the north, anticyclonic curvature of the streamlines at 700 hPa) 38 prevented it from organizing itself and propagating. The MCS1 39 and MCS2 cases were simulated using the high-resolution Méso-40 NH model and were extensively analysed using observations acquired during the intensive observation period of AMMA 41 (Barthe et al., 2010). These two situations are therefore of great 42 interest for evaluating the performance of the AROME model. It 43 should be noted that Bain et al. (2011) made a detailed analysis 44 of the structure of this AEW during the 25–29 July period. 45

46 Configuration of models, design of simulations and method 2.2. 47

48 2.2.1. The global ARPEGE forecasting system 49

50 In order to explore the sensitivity of the AROME model to initial 51 and forcing atmospheric fields provided by the French global 52 Action des Recherche Petite Échelle et Grande Échelle (ARPEGE) 53 forecasting system (Courtier *et al.*, 1994), we first briefly describe 54 its characteristics for the two versions Cycle 32 and 33 (CY32 and 33 hereafter) used in the present article. The CY32 (CY32) has a T358 (T538) spectral truncation on a stretched sph 55 56 stretching factor of 2.4, and 46 (60) vertical levels from 17 to 45 km. 57 This configuration leads to a horizontal resolution of about 50 km 58 (25 km) over West Africa. The deep convection parametrization is 59 based on the convergence of humidity (Bougeault, 1985; Ducrocq 60 and Bougeault, 1995). The major changes in parametrizations 61 brought by the CY33 relative to the CY32 were the introduction of the turbulent kinetic energy (TKE) and a new scheme of shallow 62 convection based on a mass flux closure (Bechtold et al., 2001). 63 The assimilation component of ARPEGE uses a 4D-Var scheme 64

with a 6 h time window and frequency (Veersé and Thépaut, 1998; Rabier et al., 2000).

During the AMMA-SOP, considerable effort has been devoted 3 to update and reinforce the radiosonde network over West Africa 4 (Parker et al., 2008), and increase its frequency with for example 5 four soundings per day at Niamey, Tamale and Parakou (20 miles 6 away from Djougou). Bock et al. (2007) have evidenced an overall 7 dry bias in radiosondes of about 3 kg m^{-2} as compared to GPS. 8 Agusti-Panadera et al. (2010) developed a model-based method 9 to correct the above biases depending on the sonde type and 10 the altitude. They concluded that the assimilation of the whole set of AMMA radiosondes had a positive impact if corrected 11 on the European Centre for Medium-range Weather Forecasts 12 (ECMWF) forecasts, results confirmed by Faccani et al. (2009) 13 for the ARPEGE system. The ECMWF correction method has 14 been implemented in ARPEGE from CY33. In the present article, 15 we assimilated all AMMA radiosondes; nevertheless they are not 16 corrected for CY32 experiments. 17

To improve the analysis, we also assimilated remote-sensing 18 microwave data from AMSU-B, by using the method developed by 19 Karbou et al. (2010a, 2010b). This sensor provides observations 20 sensitive to the air moisture from the surface to about 20 km height. The assimilation of channels 2 and 5 over land significantly 21 improved the analysis of the air humidity and temperature at 22 low levels up to 4 km including over West Africa (Karbou 23 et al., 2010b). 24

To assess the impact of the above observations on the analysis, 25 we performed a set of four assimilation experiments summarized 26 in Table 1, exploring the sensitivity to the cycle (CY32 and 27 CY33 names) and to the assimilation (or not) from AMSU-B 28 channels over continental surfaces with the suffix +A (or not). 29 These experiments were performed over a ~2-month AMMA 30 period (14 July to 11 September 2006), four times per day (0000-0600-1200-1800 UTC). 31

From this set of four assimilation experiments, we initialized 32 a set of four ARPEGE forecasts (ARP_ series) summarized in 33 Table 2. The ARPEGE forecasts were run over 6 h for the full ~2-34 month period to assess the analysis of precipitation. Nevertheless, 35 for the 23-28 July case-study we performed a set of ARPEGE 36 forecasts over 6 days starting from 23 July at 0000 UTC, to be 37 compared with AROME forecasts. 38



E Qassimilation experiments with the ARPEGE 4D-Var assimilation Table system.

Analysis	ARPEGE version	Truncation (resolution over Africa)	AMSU-B assimilation	Period of study for all analyses
CY32	Cycle 32	T358 (~50 km)	None	14 July to 14 September
CY32+A	Cycle 32	T358 (~50 km)	Channels 2 and 5 over continent	
CY33	Cycle 33	T598 (~25 km)	None	
СҮЗЗ+А	Cycle 33	T598 (~25 km)	Channels 2 and 5 over continent	



Table 2. Set of forecast experiments with the ARPEGE model.

Simulation	Model	Initial state (analysis)	Forced by	Range forecast and (period of study) for all simulations
ARP_CY32	ARPEGE	CY32	None	6 h (14 July to 14 September)
ARP_CY32+A	_	CY32+A	None	1
ARP_CY33	-	CY33	None	
ARP_CY33+A	-	CY33+A	None	
				144 h (23-28
				July)

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ast experiments with the AROME model

Simulation	Model	Initial state	Forced by		Range forecast and (period of study)
ARO_cplCY32	AROME	CY32	CY32	AROME forced by ARPEGE analysis (hindcast)	144 h
ARO_cplCY32+A	_	CY32+A	CY32+A	· · · · · ·	
ARO_cplCY33	_	CY33	CY33		
ARO_cplCY33+A	-	CY33+A	CY33+A		
					(23-28 July)
ARO_cplARP_CY32	AROME	CY32	ARP_CY32	AROME forced by ARPEGE forecast	144 h
ARO_cplARP_CY32+A	_	CY32+A	ARP_CY32+A		
ARO_cplARP_CY33	-	CY33	ARP_CY33		
ARO_cplARP_CY33+A	-	CY33+A	ARP_CY33+A		
					(23-28 July)

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15 2.2.2. The AROME model

The simulations studied in this article were performed using 17 the French high-resolution NWP regional model AROME (Seity 18 et al., 2011). AROME has been operational over France, with a 19 2.5 km horizontal resolution, since the end of 2008. AROME is 20 a non-hydrostatic regional model and takes advantage of both 21 the ARPEGE system (dynamical adiabatic core, data assimilation, 22 software) (Courtier et al., 1994; Bubnová et al., 1995) and the 23 Méso-NH research system (Lafore et al., 1998) for most of 24 its physical parametrizations (microphysics, turbulence, shallow 25 convection, surface). AROME is a bi-Fourier spectral limited-area 26 model and uses a two-time-level, semi-implicit, semi-Lagrangian discretization scheme on an A-grid. It reaches high numerical 27 efficiency with time steps greater than 60 s at 2.5 km resolution, 28 which is suitable for operational purposes. 29

The vertical grid contains 60 levels stretched from 10 m to 30 20 km with 28 levels below 3000 m. Physical parametrizations 31 include the Rapid Radiative Transfer Model (RRTM) long-wave 32 (Mlawer et al., 1997) and Fouquart and Bonnel (1980) short-33 wave radiation schemes, mixed-phase microphysics (Pinty and 34 Jabouille, 1998), the TKE scheme of Cuxart et al. (2000) and the Interactions Soil-Biosphere-Atmosphere (ISBA) surface scheme 35 (Noilhan and Mahfouf, 1996). The 2.5 km horizontal resolution 36 has been chosen over France in order to partly bypass the so-37 called 'convective grey zone' and to better represent among others 38 the orography, the physiographic data and the type of soil. At 39 this resolution, the deep convection is assumed to be explicitly 40 resolved by the model but the shallow convection requires a 41 specific parametrization of subgrid-scale processes for which the 42 Pergaud et al. (2009) scheme is used. It is a mass flux scheme 43 that parametrizes dry thermals and shallow cumuli. Since 2008, AROME represents a real benefit for French forecasters to handle 44 convective situations, which has been confirmed by objective 45 scores adapted to small scales. Nevertheless, AROME skill has 46 never been assessed over continental tropical regions such as 47 West Africa. 48

The AROME simulations were initialized at 0000 UTC on 49 23 July and were run over 6 days to simulate all scales and 50 processes that might impact our convective systems of interest, 51 from the monsoon surge to convection. The simulation domain 52 (Figure 1) covered a large part of West Africa and the Gulf of 53 Guinea $(0-22^{\circ}N, 15^{\circ}W-20^{\circ}E)$ at 5 km resolution, representing 54 696×444 grid points. The choice of the resolution was a compromise between the need for a resolution high enough to 55 explicitly represent deep convection (without a deep convection 56 scheme) and the use of a domain large enough to simulate 57 both the initiation and propagation of MCSs, and synoptic 58 features such as AEWs interacting with convection. As shown by 59 Weisman et al. (1997) for midlatitudes and by Barthe et al. (2010) 60 over West Africa, although it may be too crude to properly 61 simulate convective cells in research mode (Bryan et al., 2003), this resolution enables salient mesoscale aspects of convection 62 to be captured, obviating the need to use a convection scheme. 63 Some sensitivity experiments performed over West Africa at 64

2.5 km resolution show a better representation of the smaller convective features but they have indicated only a weak impact (not shown) on the representation of the life cycle of the largest 17 MCSs.

2.2.3. Initialization and forcing of AROME simulations

21 Over West Africa, we used the AROME model without its own 22 assimilation system, since data were too sparse over the region. 23 Atmospheric and surface initial fields were taken from the 4D-Var 24 assimilation system of the global ARPEGE model, whereas the 25 lateral boundary conditions were provided either by the ARPEGE forecast or by its analysis every 6 h. The one-way coupling was 26 performed at an hourly step so that ARPEGE forecasts were used 27 at intermediate ranges. 28

We initialized a first set of four AROME simulations forced 29 by the set of four ARPEGE analyses (ARO_cpl series) and a 30 second set of four AROME simulations forced by the set of four 31 ARPEGE forecasts (ARO_cplARP series). Table 3 summarizes our 32 simulations. 33

2.2.4. Evaluation procedure 34 35

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36 Evaluation against independent GPS observations of TCWV 37 During the SOP of AMMA 2006, six GPS ground stations were deployed over West Africa (Figure 1) at Timbuktu, 38 Gao, Niamey, Ouagadougou, Djougou and Tamale providing 39 hourly measurements of the TCWV (Bock et al., 2007). These 40 observations were not assimilated in NWP models and could 41 therefore be used for an independent assessment of our 42 simulations. Basic statistics (correlations, variance and bias) were 43 computed and presented as Taylor diagrams (Taylor, 2001). 44

45 Quantitative precipitation forecasts ARPEGE and AROME 46 simulation skills were quantitatively evaluated in terms of quan-47 titative precipitation forecasts (QPFs). However, an evaluation of 48this kind is not straightforward because of the natural large spatial 49 and temporal variability of precipitation, especially for convective 50 regimes over West Africa, and also because of the sparse density 51 of the rain-gauge network (see Fig. 2 in Ali et al. (2005)). Consequently, we used the TRMM-3B42 rainfall estimates derived 52 from a combination of microwave satellite observations (SSM/I, 53 TMI, AMSR and AMSU), geosynchronous infrared precipitation 54 estimates and monthly-accumulated rain-gauge data (Huffman 55 et al., 2007). This comprehensive dataset was previously evaluated 56 over our domain of interest (Roca et al., 2010; Jobard et al., 2011; 57 Pierre et al., 2011; Pedinotti et al., 2012; Gosset et al., 2013) and 58 provided instantaneous rain rates every 3 h at a spatial resolu-59 tion of $0.25^{\circ} \times 0.25^{\circ}$. For quantitative estimates (i.e. scores), 60 the TRMM-3B42 product and AROME/ARPEGE precipitation forecasts from our simulations were aggregated on a $0.5^{\circ} \times 0.5^{\circ}$ 61 62 grid.

To assess the ARPEGE and AROME simulations, we calculated 63 the Equitable Threat Score (ETS) against the 24 h cumulative 64 1 TRMM-3B42 rainfall estimate. ETS is a diagnostic score often used (Nuret et al., 2005) in the absence of a climatology event 2 since the hits of forecasts expected by chance are removed. In this 3 study, the ETS was calculated by taking 50 random test samples, 4 each comprising 25% random data, before averaging the 50 ETSs. 5 The scores were calculated for each rainfall pixel (TRMM-3B42 6 or forecast) with a range of thresholds for rainfall greater than 7 0.1, 1, 2, 5, 10 and 20 mm day⁻¹. The lowest threshold (0.1 mm 8 day^{-1}) gave insights into the system's ability to discriminate rainy 9 from non-rainy days, while the highest threshold corresponded 10 to events associated with MCS.

The 24 h cumulative precipitation was computed over 11 0600-0600 UTC periods, more in phase with the life cycle of 12 convective events. In order to evaluate the ARPEGE analysis, 13 the 24 h cumulative precipitation was calculated by cumulating 14 the 6 h range forecast of precipitation. The period of interest for 15 ARPEGE analysis spanned 19 July to 11 September 2006. For both 16 ARPEGE and AROME 144 h range forecasts, the 24 h cumulative 17 precipitation was computed over the period from 0600 UTC 23 18 July to 0600 UTC 28 July. For all 12 experiments reported in this 19 article (Table 2), ETSs were statistically 95% t-test significant, even for the 20 mm day⁻¹ threshold. 20

MCS tracking In order to characterize the individual convective 2.2 events, a tracking algorithm was applied to the surface 23 precipitation, using the 1 h accumulated precipitation. This 24 technique simply connects, in space and time, the continuous 25 regions where the 1 h accumulated precipitation exceeds 2 mm. 2.6 It corresponds to the first step of a recently developed (Fiolleau 27 and Roca, 2013) infrared-based cloud tracking algorithm. The 28 results presented here are only marginally dependent on the value 29 of the threshold used. With this object-oriented approach, we 30 characterized the morphology of the convective systems, at least in terms of their rainfall signature, in a way similar to the tracking 31 methodology traditionally used for cloud analysis which gave 32 access to a large number of parameters (see for instance Fiolleau, 33 2010). We focused on the trajectory of the centre of the rain 34 object, its duration and local time of genesis. 35

In order to validate the MCS tracking based on simulated rain, we also applied the tracking algorithm onto data based on satellite rainfall estimates GSMaP. The GSMaP v5 hourly instantaneous dataset (Ushio *et al.*, 2009) is regridded at the 0.25° resolution to mimic the simulations. Note that these data have been shown to produce reasonable rainfall characteristics over the domain of study using the AMMA campaign ground observations (Roca *et al.*, 2010).

43443. Evaluation of ARPEGE analysis and of its impacts

The present section focuses on the evaluation of the four ARPEGE
analysis experiments over the West African domain of AROME
(Figure 1) over the whole ~2-month period (14 July to 15)
September 2006). In a second step, we assess the impact of
ARPEGE fields on AROME simulations for selected case-studies.

51 3.1. Evaluation of the ARPEGE experiments

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53 3.1.1. TCWV
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Figure 2 shows the Taylor diagram to evaluate the skill of the set of ARPEGE assimilations (labelled 1-4) for the analysis

set of ARPEGE assimilations (labelled 1-4) for the analysis 56 of TCWV as compared with independent GPS observations at 57 six locations during summer 2006. We found that comparison 58 results depended on the location, with the best performance 59 reached at Niamey (correlation \sim 0.85) and the worst at Djougou 60 (\sim 0.55). This may be related to the quality of observed data, as the 61 Niamev area was one of the major focuses of AMMA observations 62 whereas calibration troubles for humidity were encountered with the Djougou radiosounding (Bock and Nuret, 2009). All other 63 locations exhibited intermediate performance levels (correlations 64

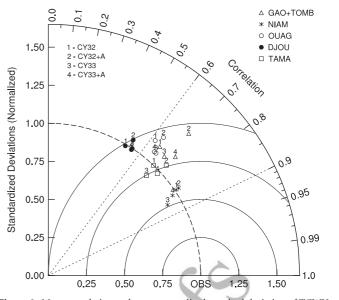


Figure 2. Mean correlation and mean normalized standard deviation of TCWV between the six GPS stations (shown on Figure 1) noted as 'OBS' and the four ARPEGE analyses (CY32, CY32+A, CY33, CY33+A) over the period 19 July to 11 September 2006. Correlations have been carried out with the nearest ARPEGE 1.5° grid point and for the four main networks (0000, 0600, 1200 and 1800 UTC).

between 0.6 and 0.75), with stronger variability at northern Sahel locations (Gao, Timbuktu, Ouagadougou). The change of cycle (32 to 33) was beneficial, with an increase of the correlation and a decrease of the variability. The assimilation of AMSU-B channels 2-5 also increased the correlation but increased the variability for both cycles. Altogether, the best performance was reached for the experiment CY33+A with the AMSU-B assimilation both for the correlation and the variability, whereas CY32 showed the smallest correlations of all GPS locations over a ~2-month period.

The ETSs for rain shown in Figure 3 are computed over a 24 h period by cumulating the ARPEGE 6 h range simulations initiated by the four ARPEGE assimilation experiments detailed in Table 1. The best scores are reached for weak to moderate rain rates $(1, 2, 5 \text{ mm day}^{-1})$, whereas heavy rainfall events appear more difficult to capture. ARP_CY33 contributes a positive impact to ETS, especially for heavy rainfalls (10 and 20 mm day⁻¹).

Note that ARP_CY33 contained a substantial modification 42 in the physics (boundary layer turbulence, deep and shallow 43 convection parametrization, radiation scheme). These changes 44 in the physics allowed a more realistic representation of the 45 boundary layer including the air moisture, low cloud, night jets 46

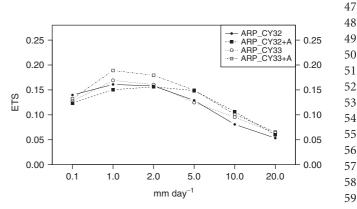


Figure 3. Impact of the assimilation of AMSUB (+A) on the Equitable Threat60Scores (ETS) of 6 h precipitation forecasts of ARPEGE for cycles CY32 and CY33,
as compared with TRMM precipitation on the same 0.5° grid for six precipitation
thresholds cumulated over a 24 h period (0600-0600 UTC). ETS are computed
over the West African domain (0-20°N, 14.8°W-20°E) of the ~2-month period
(19 July to 11 September 2006).60616162636364

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and precipitation. The assimilation of AMSU-B channels 2-5 improved the ETS with respect to ARP_CY33 and ARP_CY32 but the improvement was more significant for cycle 33 for all rain rates. A significant benefit appeared when the ARPEGE model used CY33 and assimilated observations sensitive to atmospheric moisture (from AMSU-B). This provided the best analysis (CY33+A) in terms of QPF as compared with the satellite precipitate estimate TRMM-3B42.

3.1.3. Diurnal cycle

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We now examine the ability of the best ARPEGE analysis 12 (CY33+A) to describe the diurnal cycle of TCWV as compared with GPS stations in terms of mean bias and correlation 13 (Figure 4(a) and (b)). Except for Niamey, the analysis is 14 moister than the GPS measurements (Figure 4(a)) and Djougou 15 is characterized by the largest bias, in agreement with Bock 16 and Nuret's (2009) findings. As CY33 includes the correction 17 of humidity bias of Agusti-Panadera et al. (2010), a possible 18 explanation of this positive bias might be related to the new 19 physical package of the CY33 and to the assimilation of AMSU-B. 20 Nevertheless an overestimation of the correction is not excluded. 21 The TCWV bias variability (Figure 4(a)) suggests that the 22 diurnal cycle of TCWV was modified during the analysis, with a large drop at 1200 UTC. Three terms contribute to the evolution 23 of TCWV: advection, precipitation and evaporation. Advection 24 activity is a maximum in the morning due to the nocturnal 25 jet (Couvreux et al., 2010) but cannot explain this drop. This 26 effect is combined with the effect of evaporation, which brings 27

more TCWV into the analysis. Precipitation is probably the best 1 candidate to partly explain this drop, as precipitation is too early 2 and eventually too large, which is typical of the behaviour of 3 large-scale models (see section 4). The fact that the drop is weak 4 for northern Sahel stations (Gao, Timbuktu) for which mean 5 precipitation is the smallest adds weight to this hypothesis.

6 We also looked at the diurnal variation of correlation between 7 ARPEGE analysis and GPS stations (Figure 4(b)). The comparison 8 shows a significant change of correlation depending on synoptic 9 times (between 0.55 and 0.85). Mean correlation values are in agreement with the Taylor diagram of Figure 3, with the best cor-10 relation for Niamey and the smallest for Djougou. Larger correla-11 tions are obtained at 0600 UTC when the variability is expected to 12 be the weakest. The correlation slowly decreases between 0600 and 13 1200 UTC, which may be due to first triggering of convection, but 14 its dispersion between locations is less probably due to the maxi-15 mum of data assimilated at this time. The correlation is smaller at 16 1800 UTC, except for locations in the north of the domain, and 17 is probably associated with the maximum of spatial variability of 18 TCWV generated by the maximum of convection in the afternoon. 19 During the night, the correlation increases and this is probably 20 related to the decrease of convective activity and of surface evaporation. 21

3.2. AROME sensitivity to the initial and forcing fields

Having assessed the performance of the set of assimilations, we 25 now look at their impact on the set of 6-day forecasts performed 26 between 23 and 28 July and summarized in Table 24 27

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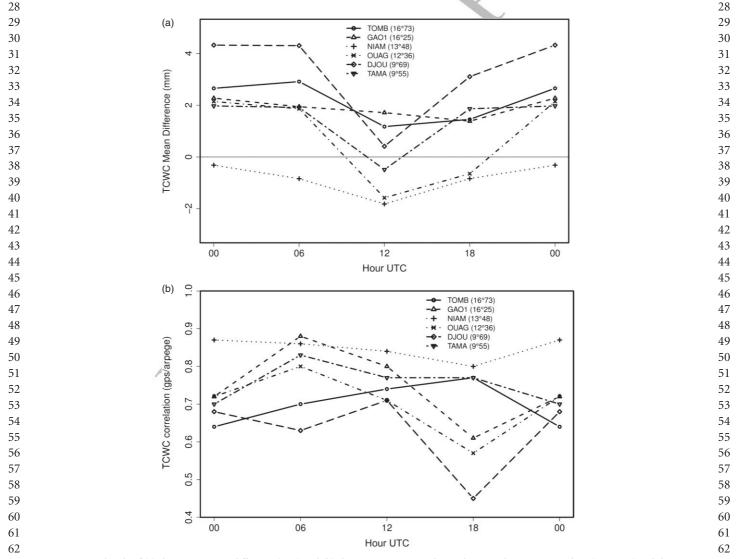


Figure 4. Diurnal cycle of (a) the mean TCWV difference (mm) and (b) the mean TCWV correlations between the ARPEGE analysis (CY33+A) and the six GPS 63 63 stations averaged over 19 July to 11 September 2006. 64 64

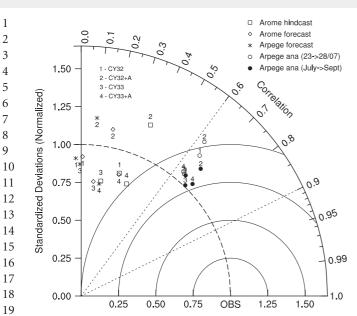


Figure 5. Taylor diagram as in Figure 2 for different sets of analyses and forecasts 20 (labelled 1-4) as compared with the ARPEGE analyses for the six GPS stations 21 over the whole 19 July to 11 September 2006 period (black dots) and over 2.2 the 23-28 July short period, for the ARPEGE analyses (open circles), for the ARPEGE forecast (black stars), and for AROME forecasts coupled with ARPEGE 23 forecast (open diamonds) and coupled with ARPEGE analyses (open squares), 24 Correlations have been calculated with the mean TCWV value of AROME in a 25 square box of 1.5° around each GPS location. 26

29 Here we evaluate the TCWV by comparing it with GPS 30 independent observations using a Taylor diagram (Figure 5). 31 First, the mean score of the ARPEGE analysis set (black dots) is 32 consistent with scores noted at the six GPS locations (Figure 2), 33 with a correlation around 0.7, a weak overestimation of the 34 variability and a positive impact of the assimilation. When a shorter period is used for the present case-study (23-28 July), 35 the results are weakly affected with just a small decrease of 36 the correlation (0.65). The skill for ARPEGE forecasts is very 37 weak or zero (starts on Figure 5) but the AMSU-B assimilation 38 improves it. AROME forecasts drastically improve the skill for 39 the TCWV, and best results are obtained when AROME is 40 coupled with the ARPEGE analysis (open squares) instead of the 41 ARPEGE forecast (open diamonds). Globally, for all experiments on the sensitivity of forecasts to initial and forcing fields, the 42 upgrade from cycle 32 to 33 (labels 1-3 and 2-4) increases 43 the correlation but decreases the variability of TCWV, whereas 44 the AMSU-B assimilation (labels 1-2 and 3-4) significantly 45 increases both the correlation and the bias (+1 to +2 mm,46 figure not shown) and weakly increases the variability of 47 TCWV. 48

49 3.2.2. QPF

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51 The precipitation ETSs of ARPEGE forecasts in this case are 52 very weak, though positive (Figure 6(a)). Nevertheless, the 53 positive impact of cycle 33 and of the assimilation of AMSU-B is noteworthy. AROME forecasts, whatever the experiment, 54 significantly improve the precipitation ETS (Figure 6(b) and 55 (c)). Their skill is even better than the ARPEGE analysis over 56 the \sim 2-month period (Figure 3). The best scores are obtained 57 when AROME is coupled with the ARPEGE analysis (Figure 6(c)) 58 instead of with the ARPEGE forecast (Figure 6(b)). The positive 59 impact of cycle 33 and the assimilation of AMSU-B on the ETS 60 is a maximum for AROME experiments coupled with ARPEGE 61 analysis (Figure 6(c)).

62 Altogether, the best analysis CY33+A allows the best AROME forecasts to be made, in terms of both TCWV and QPF, 63 especially when AROME is coupled with this analysis (experiment 64

ARO_cplCY33+A). In the following, the latter experiment will be 1 used as a reference.

4. Analysis of forecasted rain events

In this section, we study the ability of the CRM to represent the structure and life cycle of the MCSs, and the associated rainfall distribution over the Sahel.

Spatial distribution of the daily mean precipitation 4.1.

11 The average rainfall over the period from 23 to 28 July as 12 provided by AROME at 5 km resolution (Figure 7(a)) has a 13 large spatial variability at small scale, in agreement with surface 14 observations showing strong heterogeneities of rainfall from 2.5 to 20 mm day⁻¹. This is particularly true over the Sahelian 15 band on variations of 5 km (Lebel et al., 2000). In order to 16 reduce this natural high spatial variability, well reproduced by 17 AROME, daily precipitation was averaged at 50 km resolution 18 to allow comparison with the TRMM-3B42 rainfall product 19 (Figure 7(b) and (c)). Overall, there is good agreement between 20 the AROME forecasts and satellite precipitation estimates in 21 terms of northward penetration and intensity of the ITCZ. The 22 mean monsoon penetration as depicted by the intertropical 23 discontinuity (ITD) location (along the 15 °C dew-point as used by Klüpfel et al. (2011)) delimits the maximum northward 24 penetration of precipitation as depicted by observations on 2.5 Figure 7(c) (ARPEGE analysis and TRMM). The same feature is 26 simulated by both models. Nevertheless the penetration simulated 27 by AROME is less intense than observed and is shifted westward 28 by about 6° over north Mali instead over south Algeria (Cuesta 29 et al., 2010). Within the ITCZ, AROME simulates the two 30 observed precipitation maxima, around the reliefs of Cameroon 31 and Fouta Djalon. Nevertheless, the Fouta Djalon maximum is overestimated, particularly at the coast close to Conakry. Also, 32 the relative break of the ITCZ is forecast around the longitude of 33 Greenwich instead of $\sim 8^{\circ}$ W, and the observed suppression of 34 rain over the Ivory Coast is not forecast. In contrast, the ITCZ 35 in ARPEGE (Figure 7(d)) is clearly too broad and weak, and is 36 located too far towards the Equator as compared with the rainfall 37 TRMM product. In particular, the suppression of rain along the 38 Guinea coast is not reproduced. Guichard et al. (2010) have shown 39 that this failure is common to all models at low resolution, when 40the convection is not explicitly resolved. Contrary to AROME, 41 ARPEGE does not capture the spatial variability of precipitation far from mountainous areas, which is related to propagating 42 MCSs strongly coupled to AEWs. This failure will be analysed in 43 the MCS tracking, section 4.4. 44

The meridional and zonal distributions of the mean daily 45 precipitation (Figure 8) allow a quantitative comparison of 46 simulations with the TRMM estimates. AROME correctly repro-47 duces the ITCZ location (8–12°N) and shape but overestimates 48 its rainfall (maximum of 9 instead of 6 mm day⁻¹). ARPEGE 49 performs better for the intensity but the ITCZ location is shifted 50 to the south ($\sim 7^{\circ}$ N). The zonal distribution of daily precipitation averaged over the $6-20^{\circ}$ N band exhibits three regions. 51 East of 6°E AROME is in agreement with TRMM precipitation 52 (~4.5 mm day⁻¹) whereas ARPEGE slightly underestimates it 53 $(\sim 3.5 \text{ mm day}^{-1})$. In the central Niamey region $(2^{\circ}W-6^{\circ}E)$ both 54 AROME and ARPEGE weakly underestimate it (~3 instead of 55 ~4.5 mm day⁻¹). Going westward from 6° W, both simulations 56 clearly diverge from the satellite estimate. The overestimation of 57 rain around Fouta Djalon is greater for AROME. 58

4.2. Representation of MCSs, AEW and the monsoon surge

61 Similar to the analysis by Barthe et al. (2010), Figure 9 shows the longitude (a) and latitude (b) Hovmüller diagrams of the 62 3-hourly 3B42 instantaneous precipitation and of the circulation 63 as analysed by ARPEGE. During the 23-28 July period, the 64

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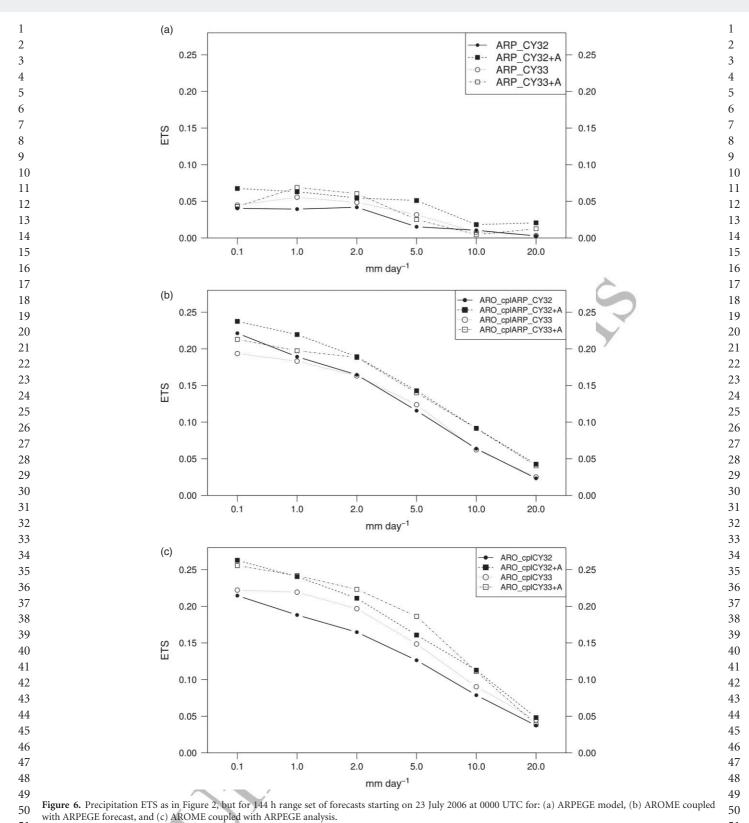
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52 convective activity was modulated by two successive major events, 53 an AEW circulation (a) followed by a monsoon surge (b). This 54 period was characterized by high AEW activity (Janicot et al., 55 2008; Bain et al., 2011) moving at a mean speed of 10 m s⁻ 56 with an intense oscillation of the meridional wind $(\pm 5\mbox{ m s}^{-1})$ 57 at 700 hPa with a \sim 3.5 day period at Niamey (Figure 9(a)). 58 Barthe et al. (2010) have shown that a monsoon surge occurred 59 on 25 and 26 July, corresponding to a northward penetration 60 of the TCWV, preceded by a reinforcement of the heat-low that 61 allowed the southwesterly monsoon flux to intensify at low level (Figure 9(b)). Convection was characterized by westward fast-62 moving systems, with the passage over Niamey of two major 63 events, MCS1 and MCS2, ahead of and in phase with the trough 64

at 0600 UTC and 2400 UTC, respectively, on 25 July. On 26–27 July, in the ridge vicinity, convection was isolated and weak before the passage of a new MCS on the following night. The latitude–time diagram (Figure 9(b)) indicates that Niamey was located just on the northern flank of MCS1 and on the southern flank of MCS2. The trajectory of the last MCS was more to the south. 53

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The AROME forecast scenario (Figure 10) shows many 59 similarities with the observed one summarized above (Figure 9): 60 similar AEW and monsoon surge, associated with westward 61 fast-moving MCSs. We can nevertheless see some differences: simulated MCS1 and MCS2 are both shifted a little to the south ($\sim 2^{\circ}$); they pass over Niamey a few hours earlier and 64

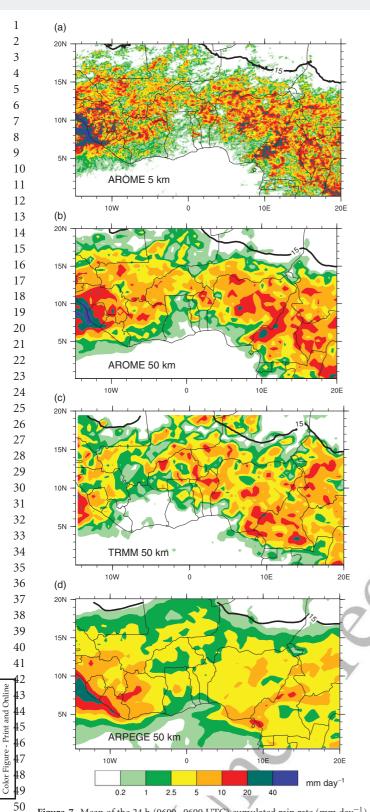


Figure 7. Mean of the 24 h (0600-0600 UTC) cumulated rain rate (mm day⁻¹)
and the surface 15 °C dew-point (mean extent of the ITD) averaged over the
23-28 July period: for the AROME forecast (a) at 0.05° resolution and (b) at
0.5° resolution; (c) for the TRMM-3B42 product at 0.5° resolution with ITD
calculated from the ARPEGE analysis at 0.5° resolution; and (d) for the ARPEGE
forecast at 0.5° resolution.

56 the diurnal convection is a little too active, especially on 26 57 July. West of 4°W, the intensification of MCSs is too high 58 compared with the TRMM-3B42 estimate, especially on arriving 59 over Fouta Djalon and at the coast. The coupling with ARPEGE, 60 which also presents too much convection in this region, may 61 have contributed to this overestimation. Nevertheless, positive 62 AEW-convection feedback could also explain this increase, since the vorticity anomalies at 700 hPa from AROME are increased 63 by a factor 5 when arriving over Fouta Djalon (not shown). 64

The same behaviour was noted for the experiment coupled with
ARPEGE (ARO_cplARP_CY33+A), while ARPEGE itself did not
exhibit such behaviour. Further analysis of AROME simulations
will be necessary to quantify and better understand the impact of
the positive feedback between MCSs and AEW, and of the Fouta
Djalon mountains (Berry and Thorncroft, 2005).1

6 Figure 11 is identical to Figures 9 and 10 except for the ARPEGE 7 forecast. Whereas the mean daily precipitation of ARPEGE was 8 fairly acceptable (Figure 8), its temporal and spatial distributions 9 at finer scale disagree markedly with observations. ARPEGE cannot capture the modulation of precipitation by the AEW 10 passage and by the monsoon surge. Also, there is no trace of 11 the fast-moving MCSs, which are a major characteristic of the 12 convection organization over the Sahel. The signal is dominated 13 by a strong diurnal cycle with a peak of convection at around 14 1500 UTC, about 3 h earlier than actually observed. Outside 15 this diurnal cycle, the convective activity depends weakly on the 16 latitude and longitude, is similar from one day to another, and 17 the period of suppressed convection behind the trough is not 18 forecast. So ARPEGE does not represent the strong variability of the convection over the Sahel, unlike AROME, which succeeds 19 thanks to its high resolution. It should be noted that ARPEGE 20 quickly damps the AEW introduced at the initial state. The 21 lack of (or wrong) coupling between convection and AEW may 2.2 provide an explanation for this behaviour of ARPEGE using a 23 convection parametrization. Sane et al. (2012) detected similar 24 bias concerning the diurnal cycle as compared with rain-gauge 25 data over Senegal for the LMDZ model. This behaviour was 26 improved by a new parametrization set.

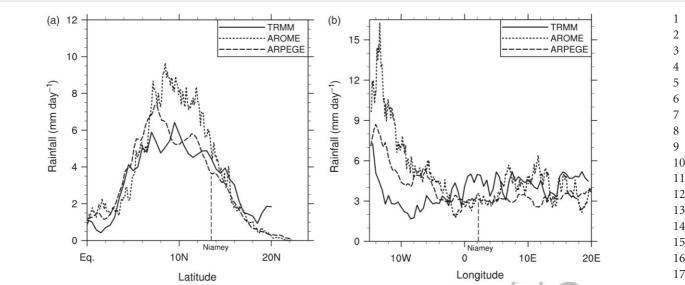
27 According to theoretical considerations (Thorncroft and Hoskins, 1994a, 1994b; Hall et al., 2006), convection is enhanced 28 ahead of or along the AEW trough. Assuming that ARPEGE-29 analysed winds are reasonably accurate, the present case-study 30 confirms this conceptual model since, at Niamey, the peaks of 31 TRMM-3B42 precipitation are located in the 700 hPa northerly 32 flow (25 July at 0900 UTC) and along the trough (26 July at 33 0000 UTC) and then decrease until 27 July in the 700 hPa 34 southerly flow (Figure 9(a)). AROME correctly simulates this 35 feature (Figure 10(a)). Thorncroft et al. (2008) showed that the location of long-lived convection is consistent with the 36 adiabatic dynamical forcing implied by the advection of the 37 vorticity perturbation ahead of AEWs. AROME (Figure 12(a)) is 38 in agreement on this point, especially for MCS1, which coincides 39 with a positive potential vorticity anomaly at 320 K (between 700 40 and 600 hPa in July over West Africa), located ahead of the AEW 41 trough.

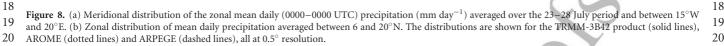
42 More recently Poan et al. (2012) showed that AEWs could be 43 also tracked by their signature in terms of TCWV and could be 44 associated with a monsoon surge also detected by a northward penetration of TCWV (Couvreux et al., 2010). Figure 12 provides 45 this 'TCWV view' for AROME using time-longitude and 46 latitude-time diagrams of the anomaly of TCWV. As suggested by 47 Poan et al. (2012), this period of high AEW activity is associated 48 with the passage of a TCWV positive anomaly (>4 mm) moving 49 at the same speed (Figures 12(a) and 10(a)) and with the 50 monsoon surge (Figure 12(b)). At finer scales, AROME shows 51 that rain events are in phase with TCWV local positive anomalies 52 (Figures 10(b) and 12(b)). This is the case for MCS1 and MCS2 (>6 mm) at their pa over Niamey and also for a first event 53 of low intensity on 24 July at 0900 UTC, which is in phase with 54 the first rain peak detected by TRMM-3B42 (Figure 9(b)) 55 strong positive TCWV anomaly in the late afternoon of 26July 56 located north of 12°N is not associated with an MCS since the 57 synoptic environment is unfavourable as analysed in detail by 58 Barthe et al. (2010). In particular, the strong southerly winds at 59 700 hPa behind the trough (Figure 9(a)) are generally associated 60 with the subsiding phase of the AEW. These features exhibited by 61 the AROME forecast are in agreement with the ARPEGE analysis. They confirm that TCWV positive anomalies favour convection 62 but other ingredients, such as the dynamics and related forcing, 63 CAPE, CIN and shear, must also be considered. 64

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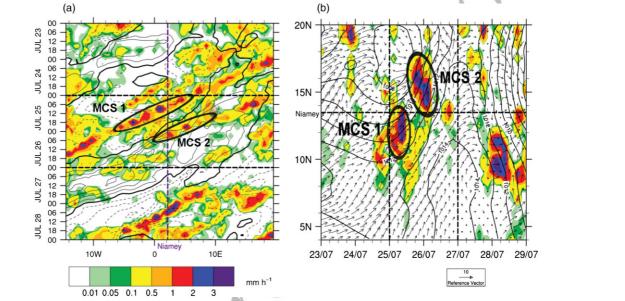


Figure 9. (a) Hovmüller diagram of the TRMM-3B42 3 h precipitation (mm h^{-1} , coloured areas) and the meridional wind at 700 hPa (isocontours every 2 m s^{-1} with a thick solid line for 0; solid and dashed lines for southerly and northerly winds respectively) for the 23-28 July period and averaged between 8 and 15°N. The horizontal lines mark the period of interest (25-26 July) and the vertical line the longitude of Niamey. (b) Similar Hovmüller diagram to (a) but for the time-latitude of TRMM-3B42 3 h precipitation (mm h⁻¹, coloured areas), mean-sea-level pressure (MSL in hPa, black lines) and wind at 925 hPa (m s⁻¹, vectors) averaged between 0 and 4°E. The horizontal dark dashed line represents the latitude of Niamey. For both figures, the black ellipses delineate rain related to MCS1 and MCS2. MSL and winds are from the ARPEGE analysis at a 6 h – 0.25° resolution and the ARPEGE forecast at a 3 h – 0.25° resolution respectively.

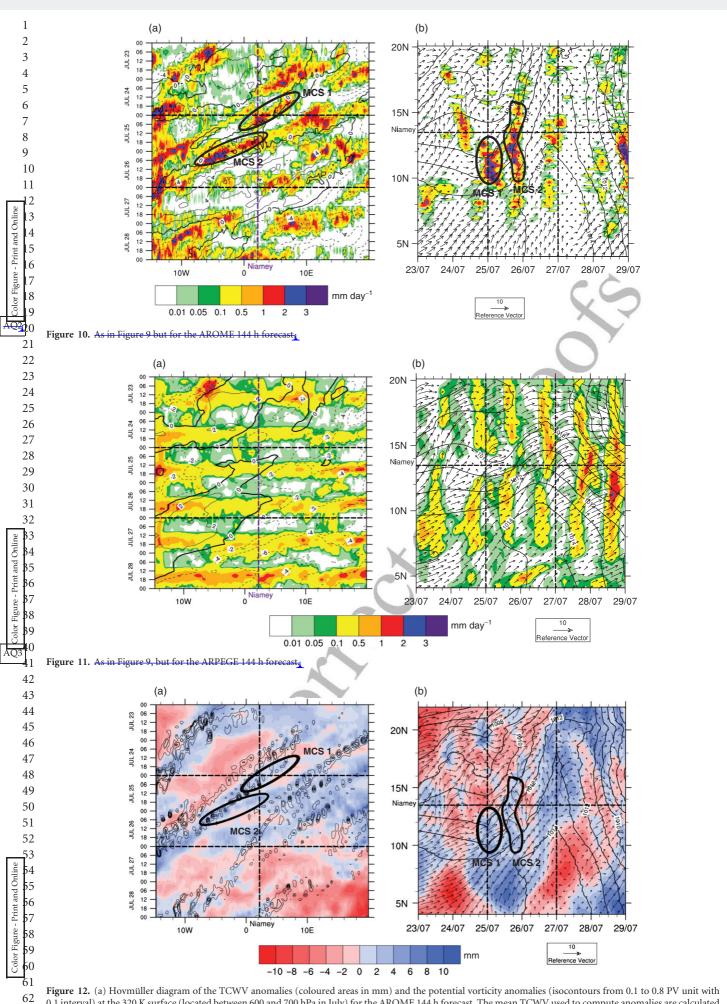
4.3. Evaluation near the Niamey AMMA supersite

The AMMA SOP in 2006 provided a unique observation dataset allowing detailed evaluation of simulations in the spirit of the one conducted by Barthe et al. (2010) for the Méso-NH research simulation system. Here, we perform an evaluation for the AROME reference simulation near the Niamey supersite. For this purpose, Figure 13 shows the temporal evolution of the vertical profile of the wind and of some surface parameters over the 23–28 July period, averaged in a 2° square box as outlined on Figure 1 (white dashed square).

AROME simulates the passage of four precipitation events through the Niamey box (blue bars in Figure 13(b)). The first event occurs on 24 July between 0900 and 1200 UTC. This system appears in TRMM-3B42 (Figure 9(b)) but the intensity of rainfall and the extent of the storm are overestimated by the model. In AROME, MCS1 passes just south of the Niamey box (Figure 10(b)) and, consequently, the event it not detected in the rainfall time series in Figure 13(b) and (c). The passage of

MCS2 is correctly simulated but 6 h early. Weak precipitation occurs on the evening of 27 July in agreement with TRMM-3B42 accumulated rain (Figure 13(c)). Finally a fourth event occurs on the evening of 28 July and is well captured by AROME. So, because of the high spatial variability of convection, the scenario of precipitation events is not recovered in detail, but the total amount of precipitation accumulated during these six days is seen similarly by TRMM-3B42 estimates and by AROME (22 mm). Barthe et al. (2010) have provided independent estimations of accumulated rain during the two days (25–26 July) in the $1^{\circ} \times 1^{\circ}$ square around Niamey computed from the AMMA-Coupling the Tropical Atmosphere and the Hydrological Cycle (AMMA-CATCH) network of 54 rain-gauges (Cappelaere et al., 2009) and from the MIT radar. They converge on a 14-15 mm value very similar to the TRMM estimate (Figure 13(c)) for the same period but on a larger, $2^{\circ} \times 2^{\circ}$, square.

The red curve on Figure 13(b) corresponds to the spatial variance of the surface wind in the Niamey box. As used by Kocha et al. (2013), it is a good indicator of the occurrence of gust fronts



616162Figure 12. (a) Hovmüller diagram of the TCWV anomalies (coloured areas in mm) and the potential vorticity anomalies (isocontours from 0.1 to 0.8 PV unit with
0.1 interval) at the 320 K surface (located between 600 and 700 hPa in July) for the AROME 144 h forecast. The mean TCWV used to compute anomalies are calculated
over the 23–29 July period for each point of longitude. (b) As in Figure 9(b) but for TCWV anomalies instead of precipitation calculated for each latitude for the
AROME 144 h forecast.6162636464

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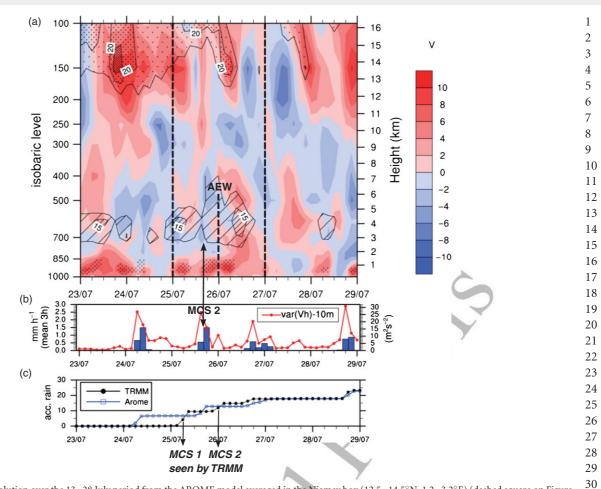
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30 Figure 13. Temporal evolution over the 13-28 July period from the AROME model averaged in the Niamey box (12.5-14.5°N, 1.2-3.2°E) (dashed square on Figure 31 1) of different parameters. (a) The meridional wind v (m s⁻¹; red and blue colours for southerly and northerly winds respectively). At low levels, periods of intense 31 monsoon flux have been outlined with dotted areas for horizontal wind speed greater than 7.5 m s^{-1} , at midlevels coarse and fine hatched areas for wind speeds 32 32 greater than 12.5 and 15 m s⁻¹ outline a strong African Easterly Jet, and at upper levels coarse, regular and fine dotted areas for wind speeds greater than 17.5, 20 and 33 33 22.5 m s⁻¹ outline a strong Tropical Easterly Jet. The thick dashed line indicates the African Easterly Wave trough as detected by the UHF radar (Fig. 6 of Barthe et al. (2010)). (b) The 3 h precipitation (mm h⁻¹) (blue histogram) and the surface wind spatial variance simulated by AROME at the Niamey box scale (m² s⁻²; red 34 34 curve). (c) The cumulated rain (mm) simulated by AROME (blue squared curve) and TRMM (3B42) estimate (dark dotted curve). 35 35

associated with convective cold pools. So, the passage of each simulated event at Niamey is associated with a peak of variance $(25-30 \text{ m}^2 \text{ s}^{-2})$ and a decrease over several hours, corresponding to the passage of the convective and stratiform parts, respectively, of the MCS as displayed by the MIT radar.

The evolution of the profile of meridional wind (Figure 13(a)) 42 can be directly compared with the observations of the UHF 43 profiler and the MIT radar at Niamey up to 6 km, as displayed by 44 Barthe et al. (2010) (their Figs 6 and 7). AROME (Figure 13(a)) 45 reproduces the observed evolution of the meridional wind 46 profile quite well (their Figs 6(b) and 7(b)). At low level, 47 the variability of monsoon flow intensity and depth fits the 48 observations, with the superposition of several scales. First, a 49 clear diurnal cycle is noted, maximum (dotted areas for winds 50 \geq 7.5 m s⁻¹ at 925 hPa) just before sunset in connection with the nocturnal jet, in agreement with previous studies (Parker et al., 51 2005; Peyrillé and Lafore, 2007; Lothon et al., 2008). Second, 52 synoptic features such as the monsoon surge and the trough 53 passage correspond to a reinforcing of the monsoon on 25 54 and 26 July, erasing its diurnal cycle. The previous heat-low 55 reinforcement on 24 July, noted on Figure 12(b), may explain the 56 monsoon intensification as suggested by Couvreux et al. (2010) 57 and Poan et al. (2012). Nevertheless, at a smaller scale, the idea 58 that convection itself contributes to this acceleration cannot be 59 excluded, in agreement with LeMone et al. (1998), who show 60 that the monsoon acceleration is maximum just ahead of MCSs and that it can occur as far as 500 km from the convective 61 62 area.

37 700 hPa, corresponding to the trough passage, observed at 38 Niamey on 26July 0000 UTC is in good agreement with the 39 UHF profiler (vertical dashed line on Figure 13(a)) (see Fig. 6b of Barthe et al. (2010)). The midlevel penetration of southerly 40 wind (up to \sim 7 km) simulated by AROME behind the AEW 41 trough resembles the observed one, but its duration and intensity 42 are a little less. The modulation of the African easterly jet (AEJ) 43 intensity around 600 hPa (hatched areas on Figure 13(a)) suggests 44 that the AEJ weakens during the passage of MCSs ($\sim 12.5 \text{ m s}^{-1}$), 45 but is reinforced ahead and behind (up to 17 m s^{-1}) in relation 46 with the Rear Inflow Jet (Smull and Houze, 1987; Lafore and 47 Moncrieff, 1989; Diongue et al., 2002). At the upper level, the 48circulation cannot be compared with observations but agrees with 49 the ARPEGE analysis (not shown), with a strong northeasterly Tropical E y Jet (TEJ) a few days before and ahead of the MCS1 and 2 and a weakening behind. The passage of these 50 51 MCSs on 25July corresponds to an intense divergent anticyclonic 52 flow resulting in an intensification of the TEJ (30 m $\ensuremath{\text{s}}^{-1})$ in 53 the southwestern sector of the convection (between 300 and 54 700 km) (figure not shown) and a reduction ($<17.5 \text{ m s}^{-1}$ at 55 150 hPa) around and at the rear of the MCS2 (up to 400 km). 56 An examination of the 150 hPa wind speed Hovmüller diagram 57 indicates that the core of the TEJ propagates westwards, staying 58 in phase with the MCS. 59

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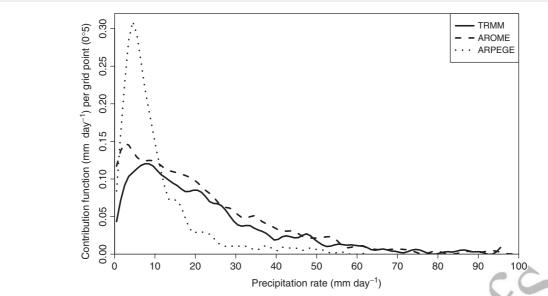
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4.4. Distribution and tracking of rain events

The intensity of rain events has a major impact on surface and 62 hydrological processes such as runoff, infiltration, soil erosion 63 and floods, especially over the Sahel, where intense showers can 64

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19Figure 14. Daily (0000-0000 UTC) rain contribution function for the TRMM-3B42 product (black curve), the AROME 144 h forecast (dashed curve), and ARPEGE206 h forecast (dotted curve) for a common 0.5° grid and by class of 1 mm day⁻¹ ranging from 0 to 100 mm day⁻¹. Data are averaged over the 23-27 July 2006 period2121

2.2 have dramatic impacts. Figure 14 displays the distribution of 23 the rain contribution to the daily precipitation intensity. To 24 allow comparison between simulations at different resolutions 25 with the TRMM-3B42 product, a common spatial (0.5°) and 2.6 temporal (daily) resolution has been chosen. AROME correctly 27 reproduces the distribution provided by satellite estimates, with 28 a tendency to overestimate all regimes. In contrast, ARPEGE 29 overestimates the number of low rain rates that are associated 30 with isolated convection, whereas it underestimates the number 31 of intense rain rate characteristics above the Sahel. These results are confirmed by the median cumulative rainfall: the AROME 32 median fits the TRMM one (17 mm day⁻¹) while the ARPEGE 33 median is largely underestimated (7 mm day⁻¹). The integral of 34 each curve gives the daily mean precipitation. As already seen on 35 Figure 8, AROME overestimates rainfall $(4.3 \text{ mm day}^{-1})$ by about 36 one quarter (+26%) as compared with TRMM $(3.4 \text{ mm day}^{-1})$, 37 whereas ARPEGE mean rainfall is correct (3.3 mm day⁻¹) at this 38 scale.

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39 To better evaluate the AROME and ARPEGE skills in 40 forecasting the MCSs, we now compare the tracking method 41 based on simulated rain with the GSMaP satellite-based rainfall 42 estimates, as explained in section 2.2. Figure 15(a)-(c) display the trajectories of MCSs lasting more than 6 h over the 25-27 43 July 2006 period, with continental initiations. The tracking from 44 the AROME simulated rain (Figure 15(b)) is in good agreement 45 with the one from the GSMaP rainfall (Figure 15(a)) in terms 46 of location, speed, life span and number of systems. In both 47 trackings, the MCS1 and MCS2 trajectories passing over Niamey 48 at the beginning and the end of 25 July respectively are captured. 49 The simulated and the 'observed' trajectories of the MCS2 are 50 very similar and cover a distance of about 2000 km from west of 51 Lake Chad to east of Guinea within about 50 h. The simulated 52 trajectory of the MCS1 passes a little too far south at Niamey as compared to the 'observed' trajectory, as already shown by 53 the TRMM-3B42 satellite-based rainfall (Figure 10(b)). Long-54 lived and fast-moving MCSs mainly occur in the Sahel band 55 $(10-15^{\circ}N)$, where the environmental conditions (CAPE, strong 56 easterly shear between the monsoon and the AEJ, midlevel dry 57 air) are favourable to the formation of intense density currents 58 that help to structure fast-moving squall-line-type MCSs (Barnes 59 and Sieckman, 1984; Rotunno et al., 1988; Lafore and Moncrieff, 60 1989). These MCSs can cover great distances and move westward. 61 In contrast, to the south, MCSs move more slowly and have a 62 shorter duration, so few MCSs can last more than 6 h. These results are consistent with several studies showing that the long-lasting 63 rainfall episodes are supported by significant CAPE and vertical 64

shear (Hodges and Thorncroft, 1997; Fink *et al.*, 2006; Mohr and
Thorncroft, 2006; Thorncroft *et al.*, 2008). MCSs simulated by
ARPEGE have a shorter duration and move more slowly, so very
few can be detected on Figure 15(c).23
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26 The tracking approach provides information useful for 27 improving knowledge of the characteristics of each observed and 28 simulated MCS, which can be studied statistically to characterize 29 the forecast skill and behaviour of each model. Figure 16 illustrates 30 the mean diurnal cycle of the continental initiation and dissipation of MCSs over the period of 23–28 July 2006. The diurnal cycles 31 simulated by AROME are in good agreement with the ones 32 deduced from GSMaP rainfall estimates. When all systems are 33 considered, GSMaP and AROME exhibit a marked diurnal cycle 34 of initiation (Figure 16(a)), maximum in the afternoon and 35 minimum in the morning. In contrast, dissipation (Figure 16(b)) 36 is minimum at noon and maximum in the evening with a decrease 37 during the night. The time difference between the initiation and 38 dissipation maxima gives a crude estimation of the most frequent 39 MCS duration (\sim 6 h). When only systems lasting more than 40 6 h are considered, the number of events decreases as expected 41 (half), and the initiation occurs a little earlier in the afternoon (Figure 16(c)). The diurnal cycle of dissipation is less marked than 42 that of initiation but more dissipation occurs during the night 43 and the morning (Figure 16(d)). Mathon and Laurent (2001) 44 provided an eight-year, high-resolution climatology of Sahelian 45 MCSs based on tracking infrared images with a time resolution of 46 0.5 h and spatial resolution 5 km. Their results (their Fig. 11(a)) 47 show that MCS initiations exhibit a large diurnal cycle with a 48 maximum between 1600 and 1700 LST and that MCS dissipations 49 occur around 2100 LST. ARPEGE shows less intense rain events 50 than AROME. Their triggering occurs earlier, between 1200 and 51 1500 UTC, not later in the afternoon (Figure 16(a)), and dissipates in the afternoon (1500-1800 UTC), resulting in a shorter mean 52 duration (\sim 3 h). Events with durations longer than 6 h are rare 53 during this short period and cannot be documented in terms of 54 diurnal cycle (Figure 16(c) and (d)). 55

In short, when comparing with the satellite-based rainfall 56 estimates, it appears that ARPEGE is unable to represent the 57 high spatial and temporal variability of precipitation associated 58 with MCSs over the Sahel, although the mean rain is correctly 59 simulated. On the other hand, a CRM such as AROME correctly 60 retrieves the spatial and temporal precipitation variability, but 61 with some overestimation of its mean value. Thus, high resolution brings substantial progress in the representation of convection and 62 associated precipitation patterns over West Africa as compared 63 with models at lower resolution using a convective scheme. 64

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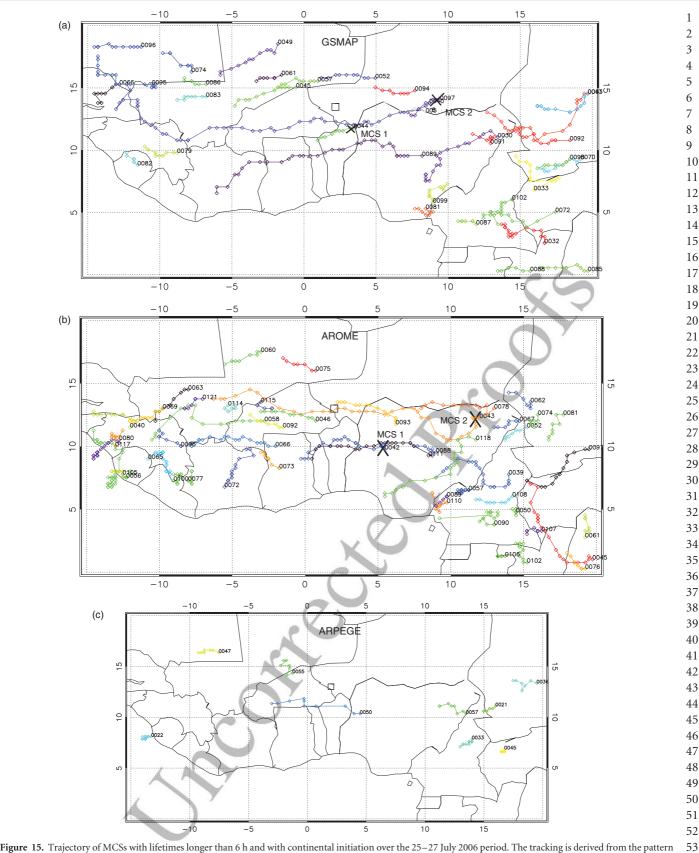


Figure 15. Trajectory of MCSs with lifetimes longer than 6 h and with continental initiation over the 25-27 July 2006 period. The tracking is derived from the pattern of rainfall heavier than 2 mm h⁻¹(a) aggregated on a 0.25° horizontal grid for satellite-based precipitation GSMaP, (b) on a 0.25° grid for the AROME simulation, and (c) on a 0.25° grid for the ARPEGE CY33+A simulation. Each diamond indicates the location of the MCS at an hourly frequency. The tracking threshold is set to 2 mm h⁻¹. The label added at the start of the trajectory corresponds to its number in the tracking file. 56

The major aim of this article was to investigate the forecasting potential of the French high-resolution NWP regional model AROME during the West African monsoon (WAM) season. We ran experiments over a large part of West Africa over a 6-day period well documented by the AMMA 2006 field campaign (Barthe *et al.*, 2010). The spatial distribution of the daily mean

precipitation shows strong similarities between the AROME forecasts and the satellite-derived precipitation TRMM-3B42, with good location of the ITCZ. The simulations produce a realistic scenario with the passage of two westward, fast-moving MCSs coupled with an intense AEW and a deep monsoon surge. AROME shows its ability to forecast long-lived MCSs and AEWs. The daily rainfall distribution shows the capacity of AROME to correctly reproduce all regimes, from light rains to the largest

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Color Figure - Print and Online

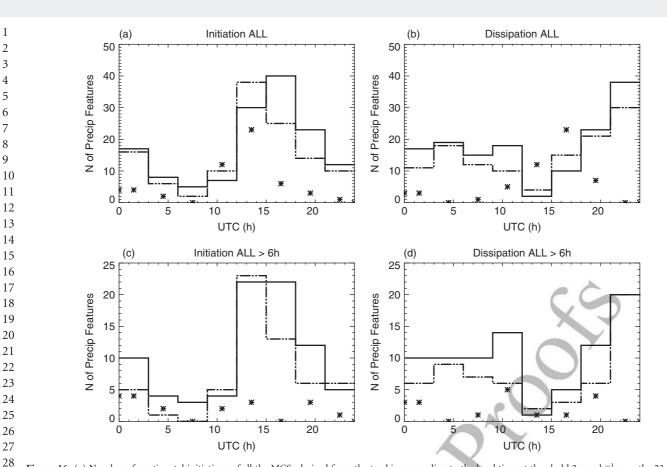


Figure 16. (a) Number of continental initiations of all the MCSs derived from the tracking according to the local time at threshold 2 mm h^{-1} over the 23–28 July 2006 period. (b) Same as (a) but for dissipation. (c) and (d) same as (a) and (b) but for MCS with lifetimes longer than 6 h. Black histograms represent AROME forecast aggregated on a 0.25° grid, dashed-dotted histograms represent the satellite-based precipitation GSMaP on a 0.25° grid and asterisks represent the CY33+A ARPEGE forecast on a 0.25° grid.

33 MCSs. A tracking approach provides useful statistics in agreement 34 with the satellite-based rainfall GSMaP and with previous studies 35 on triggering and dissipation stages, and lifetimes (Mathon et al., 36 2002). In long-lived MCSs, triggering is earlier and dissipation is 37 delayed in comparison with short-lived MCSs. This is consistent 38 with the idea of systems that are initiated earlier being more likely 39 to survive in the less favourable conditions. Whereas the ARPEGE global model correctly simulates the mean spatial distribution of 40 rain over our study period, the model cannot simulate the life cycle 41 of MCSs correctly (too short-lived and stationary), the diurnal 42 cycle of precipitation is too strong with a lack of variability 43 from one day to another, and light precipitation events are too 44 numerous in comparison to the strongest ones.

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45 This article confirms that a CRM with a horizontal resolution 46 of 5 km is a good compromise (Barthe et al., 2010) to forecast 47 the major key features of the WAM, from local to synoptic scales, 48 even if the total amount of rainfall is overestimated by about one 49 quarter. Sensitivity experiments performed at higher resolution 50 (2.5 km) show no improvement in terms of QPFs. More in-depth 51 studies are still needed to analyse the process representations and/or the numerics responsible for this overestimation, but 52 our results point to the possibly important role of the scale 53 interactions between the MCS and the AEW, which could be 54 reinforced over the Guinea mountains (Berry and Thorncroft, 55 2005; Laing et al., 2008). This rain overestimation is likely to 56 reinforce the latent heat fluxes (Guichard et al., 2010) which could 57 partly explain the overestimation of TCWV (+1 to +2 mm) and 58 further moisture supply for convection. CRM intercomparisons 59 could bring a better understanding of the shortcomings of each 60 model. For example, case-studies simulated with the COSMO 61 model over West Africa (Klüpfel et al., 2014) and with the WRF 62 model over East Africa (Laing et al., 2012) both underestimate 63 precipitation (-50 and -20% respectively) as compared with the satellite-derived precipitation. 64

33 Another objective of this article was to assess the AROME model 34 sensitivity to forcing conditions. For that aim, we assimilated 35 surface-sensitive observations from microwave remote-sensing 36 sensors AMSU-B over land surface with the ARPEGE 4D-Var 37 system. That improves significantly the ARPEGE analysis, but 38 it has a weak but positive impact on the ARPEGE forecast. On the contrary, it was found that this analysis used as initial 39 and boundary conditions significantly improves the AROME 40 forecasts, both in terms of TCWV and QPFs. The sensitivity 41 of AROME to its forced model could be partly linked with the 42 fact that, over West Africa, AROME is run without its own 43 assimilation. But it is consistent with previous studies showing 44 that the skill of the AEW forecast within different operational 45 NWP systems, such as ARPEGE, is less than 48 h. Also, Torn 46 (2010) found a strong sensitivity of the AEW forecast to the initial 47 humidity field after 1 day of forecast, whereas the impact of the 48 initial dynamics vanished quickly.

49 It appears that AROME, like other CRMs, allows a realistic simulation of weather systems over West Africa and a much better 50 representation of spatial and temporal rain distributions than 51 lower-resolution models using a convection parametrization. 52 Although some biases still remain and need further specific 53 improvements, AROME can now be used to enhance our 54 understanding of weather systems over Africa and their 55 interactions with larger-scale features such as AEWs and monsoon 56 bursts. The MCS tracking developed here from model outputs 57 appears to be a convenient tool to compare with observations and 58 to better capture some processes involved in convection.

Another key process that governs the convection over the Sahelian band is the land-surface interaction (Taylor *et al.*, 2011). 60 This coupling with the surface will be explored in a further 61 work by using surface conditions provided by the AMMA Land 62 Surface Model Intercomparison Project (ALMIP: Boone *et al.*, 63 2009). 64

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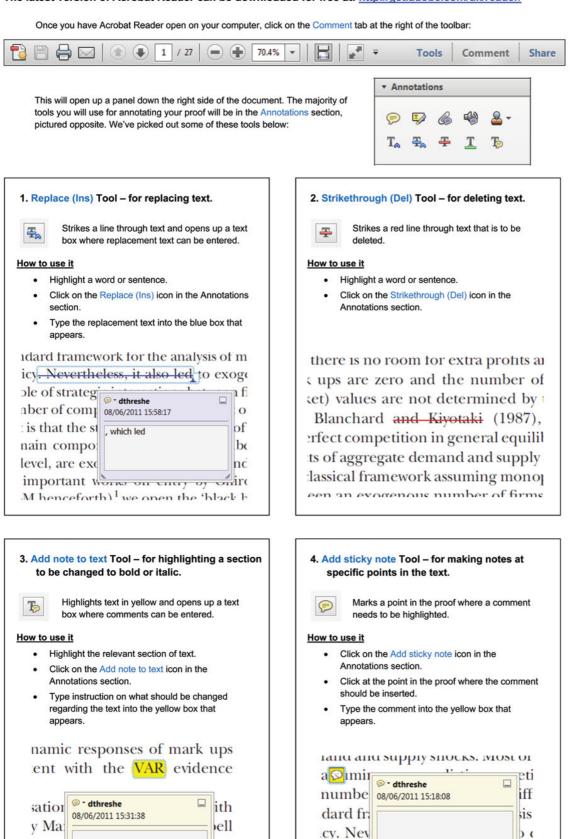
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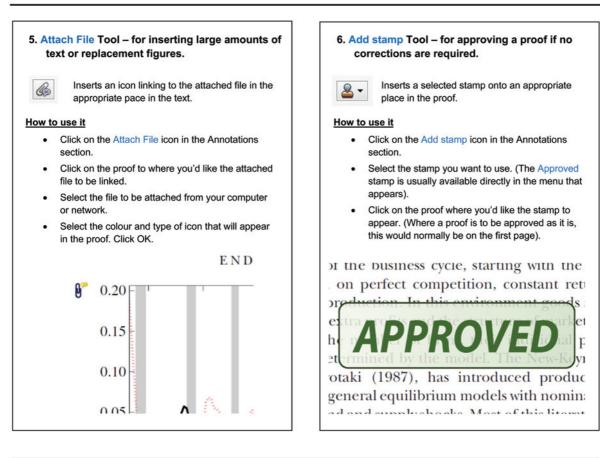
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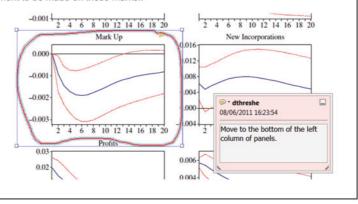


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