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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°N, 15°W–20°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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1. Introduction

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

the West African rainfall variability (West African monsoon, WAM) was one of the major aims of the African Monsoon Multidisciplinary Analysis (AMMA: <http://www.amma-international.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

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 Sahel (Reed *et al.*, 1977; Thorncroft and Hoskins, 1994a, 1994b) and significantly modulating rainfall.

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

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

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

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

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

Figure 1. Map of the AROME domain with the geographical labels (relevant orographic features appearing in italics). The grey shaded areas indicate terrain height 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 the boundaries of the domain called 'Niamey box' used in Figure 13.

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

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

2. The case-study and experimental design

2.1. Case-study

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

2.2. Configuration of models, design of simulations and method

2.2.1. The global ARPEGE forecasting system

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

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

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

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

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

Table 1. Set of assimilation experiments with the ARPEGE 4D-Var assimilation 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	
CY33+A	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	
ARP_CY33	–	CY33	None	
ARP_CY33+A	–	CY33+A	None	144 h (23–28 July)

Table 3. Set of forecast 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		
ARO_cplARP_CY32	AROME	CY32	ARP_CY32	AROME forced by ARPEGE forecast	(23–28 July) 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)

2.2.2. The AROME model

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

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

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

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 MCSs.

2.2.3. Initialization and forcing of AROME simulations

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

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

2.2.4. Evaluation procedure

Evaluation against independent GPS observations of TCWV

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

Quantitative precipitation forecasts ARPEGE and AROME simulation skills were quantitatively evaluated in terms of quantitative precipitation forecasts (QPFs). However, an evaluation of this kind is not straightforward because of the natural large spatial and temporal variability of precipitation, especially for convective regimes over West Africa, and also because of the sparse density of the rain-gauge network (see Fig. 2 in Ali *et al.* (2005)). Consequently, we used the TRMM-3B42 rainfall estimates derived from a combination of microwave satellite observations (SSM/I, TMI, AMSR and AMSU), geosynchronous infrared precipitation estimates and monthly-accumulated rain-gauge data (Huffman *et al.*, 2007). This comprehensive dataset was previously evaluated over our domain of interest (Roca *et al.*, 2010; Jobard *et al.*, 2011; Pierre *et al.*, 2011; Pedinotti *et al.*, 2012; Gosset *et al.*, 2013) and provided instantaneous rain rates every 3 h at a spatial resolution of 0.25° × 0.25°. For quantitative estimates (i.e. scores), the TRMM-3B42 product and AROME/ARPEGE precipitation forecasts from our simulations were aggregated on a 0.5° × 0.5° grid.

To assess the ARPEGE and AROME simulations, we calculated the Equitable Threat Score (ETS) against the 24 h cumulative

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

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

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

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).

3. 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.

3.1. Evaluation of the ARPEGE experiments

3.1.1. TCWV

Figure 2 shows the Taylor diagram to evaluate the skill of the set of ARPEGE assimilations (labelled 1–4) for the analysis of TCWV as compared with independent GPS observations at six locations during summer 2006. We found that comparison results depended on the location, with the best performance reached at Niamey (correlation ~0.85) and the worst at Djougou (~0.55). This may be related to the quality of observed data, as the Niamey area was one of the major focuses of AMMA observations whereas calibration troubles for humidity were encountered with the Djougou radiosounding (Bock and Nuret, 2009). All other locations exhibited intermediate performance levels (correlations

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.

3.1.2. QPF

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 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 in the physics (boundary layer turbulence, deep and shallow convection parametrization, radiation scheme). These changes in the physics allowed a more realistic representation of the boundary layer including the air moisture, low cloud, night jets

Figure 3. Impact of the assimilation of AMSUB (+A) on the Equitable Threat Scores (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).

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

We now examine the ability of the best ARPEGE analysis (CY33+A) to describe the diurnal cycle of TCWV as compared with GPS stations in terms of mean bias and correlation (Figure 4(a) and (b)). Except for Niamey, the analysis is moister than the GPS measurements (Figure 4(a)) and Djougou is characterized by the largest bias, in agreement with Bock and Nuret's (2009) findings. As CY33 includes the correction of humidity bias of Agusti-Panadera *et al.* (2010), a possible explanation of this positive bias might be related to the new physical package of the CY33 and to the assimilation of AMSU-B. Nevertheless an overestimation of the correction is not excluded.

The TCWV bias variability (Figure 4(a)) suggests that the diurnal cycle of TCWV was modified during the analysis, with a large drop at 1200 UTC. Three terms contribute to the evolution of TCWV: advection, precipitation and evaporation. Advection activity is a maximum in the morning due to the nocturnal jet (Couvreux *et al.*, 2010) but cannot explain this drop. This effect is combined with the effect of evaporation, which brings

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

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

3.2. AROME sensitivity to the initial and forcing fields

Having assessed the performance of the set of assimilations, we now look at their impact on the set of 6-day forecasts performed between 23 and 28 July and summarized in Table 2.

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 stations averaged over 19 July to 11 September 2006.

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

3.2.1. TCWV

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

3.2.2. QPF

The precipitation ETs of ARPEGE forecasts in this case are very weak, though positive (Figure 6(a)). Nevertheless, the positive impact of cycle 33 and of the assimilation of AMSU-B is noteworthy. AROME forecasts, whatever the experiment, significantly improve the precipitation ETs (Figure 6(b) and (c)). Their skill is even better than the ARPEGE analysis over the ~ 2 -month period (Figure 3). The best scores are obtained when AROME is coupled with the ARPEGE analysis (Figure 6(c)) instead of with the ARPEGE forecast (Figure 6(b)). The positive impact of cycle 33 and the assimilation of AMSU-B on the ETs is a maximum for AROME experiments coupled with ARPEGE analysis (Figure 6(c)).

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

ARO_cplCY33+A). In the following, the latter experiment will be 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.

4.1. Spatial distribution of the daily mean precipitation

The average rainfall over the period from 23 to 28 July as provided by AROME at 5 km resolution (Figure 7(a)) has a large spatial variability at small scale, in agreement with surface observations showing strong heterogeneities of rainfall from 2.5 to 20 mm day⁻¹. This is particularly true over the Sahelian band on variations of 5 km (Lebel *et al.*, 2000). In order to reduce this natural high spatial variability, well reproduced by AROME, daily precipitation was averaged at 50 km resolution to allow comparison with the TRMM-3B42 rainfall product (Figure 7(b) and (c)). Overall, there is good agreement between the AROME forecasts and satellite precipitation estimates in terms of northward penetration and intensity of the ITCZ. The mean monsoon penetration as depicted by the intertropical discontinuity (ITD) location (along the 15 °C dew-point as used by Klüpfel *et al.* (2014)) delimits the maximum northward penetration of precipitation as depicted by observations on Figure 7(c) (ARPEGE analysis and TRMM). The same feature is simulated by both models. Nevertheless the penetration simulated by AROME is less intense than observed and is shifted westward by about 6° over north Mali instead over south Algeria (Cuesta *et al.*, 2010). Within the ITCZ, AROME simulates the two observed precipitation maxima, around the reliefs of Cameroon and Fouta Djallon. Nevertheless, the Fouta Djallon maximum is overestimated, particularly at the coast close to Conakry. Also, the relative break of the ITCZ is forecast around the longitude of Greenwich instead of $\sim 8^\circ\text{W}$, and the observed suppression of rain over the Ivory Coast is not forecast. In contrast, the ITCZ in ARPEGE (Figure 7(d)) is clearly too broad and weak, and is located too far towards the Equator as compared with the rainfall TRMM product. In particular, the suppression of rain along the Guinea coast is not reproduced. Guichard *et al.* (2010) have shown that this failure is common to all models at low resolution, when the convection is not explicitly resolved. Contrary to AROME, ARPEGE does not capture the spatial variability of precipitation far from mountainous areas, which is related to propagating MCSs strongly coupled to AEWs. This failure will be analysed in the MCS tracking, section 4.4.

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

4.2. Representation of MCSs, AEW and the monsoon surge

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

Figure 6. Precipitation ETS as in Figure 2, but for 144 h range set of forecasts starting on 23 July 2006 at 0000 UTC for: (a) ARPEGE model, (b) AROME coupled with ARPEGE forecast, and (c) AROME coupled with ARPEGE analysis.

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

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.

The AROME forecast scenario (Figure 10) shows many similarities with the observed one summarized above (Figure 9): similar AEW and monsoon surge, associated with westward 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

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.

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

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 Djallon mountains (Berry and Thorncroft, 2005).

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

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

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

Figure 8. (a) Meridional distribution of the zonal mean daily (0000–0000 UTC) precipitation (mm day^{-1}) averaged over the 23–28 July period and between 15°W and 20°E . (b) Zonal distribution of mean daily precipitation averaged between 6 and 20°N . The distributions are shown for the TRMM-3B42 product (solid lines), AROME (dotted lines) and ARPEGE (dashed lines), all at 0.5° resolution.

Figure 9. (a) Hovmöller 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 Hovmöller 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 $6 \text{ h} - 0.25^{\circ}$ resolution and the ARPEGE forecast at a $3 \text{ h} - 0.25^{\circ}$ 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 is 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

Figure 10. As in Figure 9 but for the AROME 144 h forecast.

Figure 11. As in Figure 9, but for the ARPEGE 144 h forecast.

Figure 12. (a) Hovmöller 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.

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 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 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 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 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 curve). (c) The cumulated rain (mm) simulated by AROME (blue squared curve) and TRMM (3B42) estimate (dark dotted curve).

associated with convective cold pools. So, the passage of each simulated event at Niamey is associated with a peak of variance (25–30 m² 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)) can be directly compared with the observations of the UHF profiler and the MIT radar at Niamey up to 6 km, as displayed by Barthe *et al.* (2010) (their Figs 6 and 7). AROME (Figure 13(a)) reproduces the observed evolution of the meridional wind profile quite well (their Figs 6(b) and 7(b)). At low level, the variability of monsoon flow intensity and depth fits the observations, with the superposition of several scales. First, a clear diurnal cycle is noted, maximum (dotted areas for winds ≥ 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.*, 2005; Peyrillé and Lafore, 2007; Lothon *et al.*, 2008). Second, synoptic features such as the monsoon surge and the trough passage correspond to a reinforcing of the monsoon on 25 and 26 July, erasing its diurnal cycle. The previous heat-low reinforcement on 24 July, noted on Figure 12(b), may explain the monsoon intensification as suggested by Couvreur *et al.* (2010) and Poan *et al.* (2012). Nevertheless, at a smaller scale, the idea that convection itself contributes to this acceleration cannot be excluded, in agreement with LeMone *et al.* (1998), who show that the monsoon acceleration is maximum just ahead of MCSs and that it can occur as far as 500 km from the convective area.

As previously discussed with the time–longitude diagram (Figure 9(a)), the reversal of northerly to southerly winds at

700 hPa, corresponding to the trough passage, observed at Niamey on 26 July 0000 UTC is in good agreement with the UHF profiler (vertical dashed line on Figure 13(a)) (see Fig. 6b of Barthe *et al.* (2010)). The midlevel penetration of southerly wind (up to ~ 7 km) simulated by AROME behind the AEW trough resembles the observed one, but its duration and intensity are a little less. The modulation of the African easterly jet (AEJ) intensity around 600 hPa (hatched areas on Figure 13(a)) suggests that the AEJ weakens during the passage of MCSs (~ 12.5 m s⁻¹), but is reinforced ahead and behind (up to 17 m s⁻¹) in relation with the Rear Inflow Jet (Smull and Houze, 1987; Lafore and Moncrieff, 1989; Diongue *et al.*, 2002). At the upper level, the circulation cannot be compared with observations but agrees with the ARPEGE analysis (not shown), with a strong northeasterly Tropical Easterly Jet (TEJ) a few days before and ahead of the MCS1 and MCS2 and a weakening behind. The passage of these MCSs on 25 July corresponds to an intense divergent anticyclonic flow resulting in an intensification of the TEJ (30 m s⁻¹) in the southwestern sector of the convection (between 300 and 700 km) (figure not shown) and a reduction (< 17.5 m s⁻¹ at 150 hPa) around and at the rear of the MCS2 (up to 400 km). An examination of the 150 hPa wind speed Hovmöller diagram indicates that the core of the TEJ propagates westwards, staying in phase with the MCS.

4.4. Distribution and tracking of rain events

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

Figure 14. Daily (0000–0000 UTC) rain contribution function for the TRMM-3B42 product (black curve), the AROME 144 h forecast (dashed curve), and ARPEGE 6 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 period and in the $(0-20^\circ\text{N}, 15^\circ\text{W}-20^\circ\text{E})$ domain.

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

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

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).

The tracking approach provides information useful for improving knowledge of the characteristics of each observed and simulated MCS, which can be studied statistically to characterize the forecast skill and behaviour of each model. Figure 16 illustrates the mean diurnal cycle of the continental initiation and dissipation of MCSs over the period of 23–28 July 2006. The diurnal cycles simulated by AROME are in good agreement with the ones deduced from GSMaP rainfall estimates. When all systems are considered, GSMaP and AROME exhibit a marked diurnal cycle of initiation (Figure 16(a)), maximum in the afternoon and minimum in the morning. In contrast, dissipation (Figure 16(b)) is minimum at noon and maximum in the evening with a decrease during the night. The time difference between the initiation and dissipation maxima gives a crude estimation of the most frequent MCS duration ($\sim 6 \text{ h}$). When only systems lasting more than 6 h are considered, the number of events decreases as expected (half), and the initiation occurs a little earlier in the afternoon (Figure 16(c)). The diurnal cycle of dissipation is less marked than that of initiation but more dissipation occurs during the night and the morning (Figure 16(d)). Mathon and Laurent (2001) provided an eight-year, high-resolution climatology of Sahelian MCSs based on tracking infrared images with a time resolution of 0.5 h and spatial resolution 5 km. Their results (their Fig. 11(a)) show that MCS initiations exhibit a large diurnal cycle with a maximum between 1600 and 1700 LST and that MCS dissipations occur around 2100 LST. ARPEGE shows less intense rain events than AROME. Their triggering occurs earlier, between 1200 and 1500 UTC, not later in the afternoon (Figure 16(a)), and dissipates in the afternoon (1500–1800 UTC), resulting in a shorter mean duration ($\sim 3 \text{ h}$). Events with durations longer than 6 h are rare during this short period and cannot be documented in terms of diurnal cycle (Figure 16(c) and (d)).

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

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.

5. Conclusion and discussion

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

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+ARPEGE forecast on a 0.25° grid.

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

This article confirms that a CRM with a horizontal resolution of 5 km is a good compromise (Barthe *et al.*, 2010) to forecast the major key features of the WAM, from local to synoptic scales, even if the total amount of rainfall is overestimated by about one quarter. Sensitivity experiments performed at higher resolution (2.5 km) show no improvement in terms of QPFs. More in-depth studies are still needed to analyse the process representations and/or the numerics responsible for this overestimation, but our results point to the possibly important role of the scale interactions between the MCS and the AEW, which could be reinforced over the Guinea mountains (Berry and Thorncroft, 2005; Laing *et al.*, 2008). This rain overestimation is likely to reinforce the latent heat fluxes (Guichard *et al.*, 2010) which could partly explain the overestimation of TCWV (+1 to +2 mm) and further moisture supply for convection. CRM intercomparisons could bring a better understanding of the shortcomings of each model. For example, case-studies simulated with the COSMO model over West Africa (Klöpfer *et al.*, 2011) and with the WRF model over East Africa (Laing *et al.*, 2012) both underestimate precipitation (−50 and −20% respectively) as compared with the satellite-derived precipitation.

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

It appears that AROME, like other CRMs, allows a realistic simulation of weather systems over West Africa and a much better representation of spatial and temporal rain distributions than lower-resolution models using a convection parametrization. Although some biases still remain and need further specific improvements, AROME can now be used to enhance our understanding of weather systems over Africa and their interactions with larger-scale features such as AEWs and monsoon bursts. The MCS tracking developed here from model outputs appears to be a convenient tool to compare with observations and 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). This coupling with the surface will be explored in a further work by using surface conditions provided by the AMMA Land Surface Model Intercomparison Project (ALMIP: Boone *et al.*, 2009).

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standard framework for the analysis of microeconomic theory. Nevertheless, it also led to the development of a number of strategic substitutes. The number of competitors in the industry is that the structure of the industry is important. The main components of the industry are the level, are excluded. The important works on entry by Shirai (1987) henceforth. We open the 'black box'.

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there is no room for extra profits and mark-ups are zero and the number of firms (set) values are not determined by the Blanchard and Kiyotaki (1987), perfect competition in general equilibrium of aggregate demand and supply in the classical framework assuming monopolistic competition and an exogenous number of firms.

3. **Add note to text** Tool – for highlighting a section to be changed to bold or italic.

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How to use it

- Highlight the relevant section of text.
- Click on the **Add note to text** icon in the Annotations section.
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dynamic responses of mark-ups consistent with the VAR evidence.

sation by Markov. The standard framework for the analysis of microeconomic theory. Nevertheless, it also led to the development of a number of strategic substitutes. The number of competitors in the industry is that the structure of the sector is important. The main components of the industry are the level, are excluded. The important works on entry by Shirai (1987) henceforth. We open the 'black box'.

4. **Add sticky note** Tool – for making notes at specific points in the text.

Marks a point in the proof where a comment needs to be highlighted.

How to use it

- Click on the **Add sticky note** icon in the Annotations section.
- Click at the point in the proof where the comment should be inserted.
- Type the comment into the yellow box that appears.

land and supply shocks. Most of the standard framework for the analysis of microeconomic theory. Nevertheless, it also led to the development of a number of strategic substitutes. The number of competitors in the industry is that the structure of the sector is important. The main components of the industry are the level, are excluded. The important works on entry by Shirai (1987) henceforth. We open the 'black box'.

USING e-ANNOTATION TOOLS FOR ELECTRONIC PROOF CORRECTION

5. Attach File Tool – for inserting large amounts of text or replacement figures.

Inserts an icon linking to the attached file in the appropriate place in the text.

How to use it

- Click on the [Attach File](#) icon in the Annotations section.
- Click on the proof to where you'd like the attached file to be linked.
- Select the file to be attached from your computer or network.
- Select the colour and type of icon that will appear in the proof. Click OK.

6. Add stamp Tool – for approving a proof if no corrections are required.

Inserts a selected stamp onto an appropriate place in the proof.

How to use it

- Click on the [Add stamp](#) icon in the Annotations section.
- Select the stamp you want to use. (The [Approved](#) stamp is usually available directly in the menu that appears).
- Click on the proof where you'd like the stamp to appear. (Where a proof is to be approved as it is, this would normally be on the first page).

of the business cycle, starting with the
on perfect competition, constant ret
production. In this environment goods
extra costs and the cost of market
he... determined by the model. The New-Key
otaki (1987), has introduced produc
general equilibrium models with nomin
ad and supply shocks. Most of this literat

APPROVED

Drawing Markups

7. Drawing Markups Tools – for drawing shapes, lines and freeform annotations on proofs and commenting on these marks.

Allows shapes, lines and freeform annotations to be drawn on proofs and for comment to be made on these marks..

How to use it

- Click on one of the shapes in the [Drawing Markups](#) section.
- Click on the proof at the relevant point and draw the selected shape with the cursor.
- To add a comment to the drawn shape, move the cursor over the shape until an arrowhead appears.
- Double click on the shape and type any text in the red box that appears.

For further information on how to annotate proofs, click on the [Help](#) menu to reveal a list of further options: