

Evaluation of a mesoscale coupled ocean-atmosphere configuration for tropical cyclone forecasting in the South West Indian Ocean basin

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Abstract

The performance in term of tropical cyclone track and intensity prediction of the new coupled ocean-atmosphere system based on the operational atmospheric model AROME-Indian Ocean and the ocean model NEMO is assessed against that of the current operational configuration in the case of seven recent tropical cyclones. Five different configurations of the forecast system are evaluated: two with the coupled system, two with an ocean mixed layer parameterization and one with a constant sea surface temperature. For each ocean-atmosphere coupling option, one is initialized directly with the MERCATOR-Ocean PSY4 product as in the current operational configuration and the other with the ocean state that is cycled in the AROME-NEMO coupled suite since a few days before the cyclone intensification. The results show that the coupling with NEMO improves the intensity of cyclones in AROME-IO, especially when they encounter a slow propagation phase. For short-term forecasts (less than 36 hours), the presence of a cooling in the initial state that has been triggered by the AROME high-resolution cyclonic winds in a previous coupled forecast already improves the tropical cyclone intensity for all coupled or uncoupled configurations. However, the simplification of the ocean-atmosphere interactions in the configurations using the ocean mixed layer parameterization is not the only reason for the overestimation of the intensity of already well-developed TC in AROME-IO. The impact of other model components, such as the air-sea flux parameterization and the cloud microphysics scheme will need to be further investigated.

1 **Evaluation of a mesoscale coupled ocean-atmosphere**
2 **configuration for tropical cyclone forecasting in the**
3 **South West Indian Ocean basin**

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10 **Key Points:**

- 11 • The numerical weather prediction coupled ocean-atmosphere model AROME-IO/NEMO
12 improves the regional forecast of Tropical cyclone in the South West Indian Ocean
13 when compared to the current operational configuration for which the ocean is re-
14 duced to a simple 1D ocean mixed layer parameterization.
- 15 • The improvement mostly comes from quasi-stationary or very slow moving intense
16 cyclones.
- 17 • In an operational suite, the fast and small scale features which are triggered in the
18 ocean by the high resolution cyclonic winds from the mesoscale atmospheric model
19 must be cycled from one coupled forecast to the next one in order to keep a mem-
20 ory of the ocean-atmosphere interactions in the vicinity of the TC.

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Abstract

The performance in term of tropical cyclone track and intensity prediction of the new coupled ocean-atmosphere system based on the operational atmospheric model AROME-Indian Ocean and the ocean model NEMO is assessed against that of the current operational configuration in the case of seven recent tropical cyclones. Five different configurations of the forecast system are evaluated: two with the coupled system, two with an ocean mixed layer parameterization and one with a constant sea surface temperature. For each ocean-atmosphere coupling option, one is initialized directly with the MERCATOR-Ocean PSY4 product as in the current operational configuration and the other with the ocean state that is cycled in the AROME-NEMO coupled suite since a few days before the cyclone intensification. The results show that the coupling with NEMO improves the intensity of cyclones in AROME-IO, especially when they encounter a slow propagation phase. For short-term forecasts (less than 36 hours), the presence of a cooling in the initial state that has been triggered by the AROME high-resolution cyclonic winds in a previous coupled forecast already improves the tropical cyclone intensity for all coupled or uncoupled configurations. However, the simplification of the ocean-atmosphere interactions in the configurations using the ocean mixed layer parameterization is not the only reason for the overestimation of the intensity of already well-developed TC in AROME-IO. The impact of other model components, such as the air-sea flux parameterization and the cloud microphysics scheme will need to be further investigated.

Plain Language Summary

The ocean provides the energy for the intensification and persistence of tropical cyclones through warm sea surface temperature and sea-air heat and moisture exchanges. But, the ocean-atmosphere interactions also trigger processes which cools the sea surface temperature beneath the tropical cyclone and thus generates a negative feedback on the TC intensification.

The numerical forecasts of the regional numerical weather prediction model AROME-IO are valuable guidance for the Regional Specialized Meteorological Centre for Tropical Cyclones of La Réunion. Currently, AROME-IO interacts with a very simplified model of the first well mixed layer of the ocean. The ocean mixed layer model is able to reproduce the turbulent mixing near the ocean surface, but its reduced physics does not allow the larger scale horizontal and vertical transport by the currents. A coupling between

53 the atmospheric model and a full ocean model is necessary to take into account the ad-
54 vective transport responsible for the upwellings and thus represent all the processes con-
55 tributing to the cooling of the ocean during the passage of a TC. The objective of our
56 study is to evaluate the possibility of replacing the current ocean mixed layer parame-
57 terization by the ocean model NEMO in the operational configuration of AROME-IO.
58 Overall, we found that the new coupling improves the cyclone intensity in AROME-IO
59 both in terms of bias and standard deviation. These improvements come almost entirely
60 from tropical cyclones that encounter a slow propagation phase (less than 2 m s^{-1} for
61 at least 12 hours). For short-term forecasts (less than 36 hours), the presence of a cool-
62 ing that is triggered by AROME high-resolution cyclonic winds in the initial state of the
63 ocean already improves the TC intensity forecast, even when the ocean mixed layer pa-
64 rameterization is used.

65 **1 Introduction**

66 Tropical cyclones (TCs) can be associated with a devastating combination of sev-
67 eral hazardous phenomena: storm surges, floods, extreme winds, tornadoes and light-
68 nings. They are highly destructive atmospheric phenomena that cause damage to life and
69 infrastructures along tropical coastal areas, particularly in the South West Indian Ocean
70 (SWIO: $30\text{-}90^\circ \text{ E}$, $0\text{-}40^\circ \text{ S}$) tropical islands, where economic vulnerability, fragile infras-
71 tructures and confined space make populations highly vulnerable to cyclonic hazards.
72 The SWIO accounts for an average of 12% of annual global cyclone activity with about
73 ten tropical storms, four of which reach the stage of a tropical cyclone (equivalent CAT2-
74 3 on the US Saffir-Simpson scale, see Leroux et al., 2018). Forecasters of the Regional
75 Specialised Meteorological Centre for Tropical Cyclones (RSMC-Cyclones) of La Réunion
76 are in demand of accurate forecasts of the track and intensity of TC to quickly identify
77 potentially impacted areas and effectively warn communities of the impending danger.

78 The work carried out in recent years has particularly led to improvements in the
79 forecasting of TC tracks (Courtney et al., 2019; Heming et al., 2019). TC intensity is less
80 predictable as it involves small scales features which are not yet well understood. The
81 implementation of coupled ocean-atmosphere (OA) numerical weather prediction (NWP)
82 systems has however been recognised as one of the key elements of the progress for TC
83 intensity forecasting (Yablonsky, 2016; Mogensen et al., 2017; Feng et al., 2019; Vellinga
84 et al., 2020). Accurate modelling of OA interactions is particularly crucial in the SWIO

85 basin, where atmospheric variability is associated with a particularly strong oceanic re-
86 sponse (and vice versa). It is considered as the cyclonic basin with the highest preva-
87 lence of OA interactions (Vialard et al., 2008) due to the unique thermocline structure
88 in the Seychelles-Chagos thermocline ridge area (55-70°E, 5-15°S) (Hermes & Reason,
89 2009).

90 The ocean provides the energy for the intensification and persistence of TCs through
91 warm sea surface temperature (SST) and air-sea heat and moisture fluxes. But, the OA
92 interactions also trigger processes which cools the SST beneath the TC and thus gen-
93 erates a negative feedback on the TC intensification (Srinivas et al., 2016).

94 The sea surface cooling is governed by three different processes: 1) loss of energy
95 associated to the heat and evaporation fluxes towards the atmosphere, 2) turbulent ver-
96 tical mixing in the upper ocean which incorporates colder water at the bottom of the
97 ocean mixed layer (OML), and 3) upwelling of deeper cold water generated by Ekman
98 pumping (Price, 1981; Black, 1983; Bender et al., 1993). The respective part of these dif-
99 ferent processes depend largely on the cyclone intensity and ocean preconditioning (Vincent
100 et al., 2012b). According to Vincent et al. (2012b) and Jullien et al. (2012), the surface
101 heat fluxes are dominant for weak to medium intensity cyclones, while the turbulent mix-
102 ing accounts for 30 to 50% of the surface cooling for weaker cyclones but for more than
103 80% for the most intense. The effect of the upwelling of colder water increases with the
104 cyclone intensity to reach 20% for the most intense cyclones. Especially, its effect is cru-
105 cial to explain the asymmetry of the cold wake with respect to the cyclone direction. By
106 not coupling the atmosphere with the ocean, the ocean acts as an infinite energy source
107 for the TCs. Coupled OA models introduce a negative feedback between the TC and the
108 SST. Although the effect of the surface heat fluxes and turbulent mixing can be repre-
109 sented by a one-dimensional (1D) parameterization of the OML, the effect of the upwellings
110 can only be represented by a three-dimensional (3D) ocean model (Yablonsky & Ginis,
111 2009; Mogensen et al., 2017). Numerous studies have thus shown the benefit of account-
112 ing for the upwelling effect in a 3D ocean model for TC forecasting (Ginis, 2002; Ben-
113 der et al., 2007; Yablonsky & Ginis, 2009; Jullien et al., 2014; Mogensen et al., 2017; Bielli
114 et al., 2021).

115 The strength of the negative feedback between the TC and the ocean and the cool-
116 ing of the surface waters is influenced by different factors; the slow translation speed and

117 shallow depth of the OML appear to favour greater SST cooling and negative feedback
118 (Schade & Emanuel, 1999). Using a simple coupled model, Schade and Emanuel (1999)
119 suggest a range of 10-60% decrease in storm intensity as a function of storm speed. Sim-
120 ilarly, Yablonsky and Ginis (2009) demonstrate the importance of a 3D ocean model for
121 resolving SST cooling when simulating a TC moving at a speed of 5 m s^{-1} and even more
122 critical for slower systems. Other subsurface ocean features such as the thermocline strat-
123 ification due to salt layers for instance or fine scales structures associated with fronts and
124 eddies also have the potential to affect storm intensity (Schade & Emanuel, 1999; Bao
125 et al., 2000; Lin et al., 2008; Vincent et al., 2012a; Jullien et al., 2014). For example, the
126 deeper mixed layer of warm-core eddies is known to favour TC intensification (Chang
127 & Anthes, 1978; Bao et al., 2000).

128 The mesoscale numerical weather prediction model "Applications de la Recherche
129 à l'Opérationnel à Méso-Echelle - Indian Ocean" (AROME-IO) provides 4 times a day
130 a forecast for a large SWIO domain. During the SWIO TC season, AROME-IO prod-
131 ucts provide valuable information on TC intensity to the RSMC-Cyclones of La Réunion.
132 In its current operational version, AROME-IO is coupled every 5 minutes to a 1D pa-
133 rameterization of the OML that takes into account the rapid change of the SST due to
134 the OA heat exchanges and the turbulent mixing of the OML that is triggered by cy-
135 clonic winds.

136 The objective of this study is to analyse the impact of replacing the current 1D OML
137 parameterization by a fully 3D coupling with the ocean model NEMO (The Nucleus for
138 European Model of the Ocean, Madec et al., 2019) in the operational high resolution mesoscale
139 cloud model AROME-IO (Seity et al., 2011; Bousquet et al., 2020) used by the RSMC
140 of La Réunion. This paper does not aim at analysing in detail the OA processes in TC
141 case studies as the current knowledge of the main OA interaction mechanisms in a TC
142 is already quite advanced and well documented. It rather focuses on the impact of chang-
143 ing the representation of the ocean component in the forecast system in term of TC fore-
144 cast. Thus, five configurations of the AROME-IO modelling system has been set up for
145 this study. Numerical simulations have been performed for a selection of seven TCs that
146 developed over the SWIO basin during the 2018-2019 and 2019-2020 cyclone seasons, and
147 the TC Batsirai, which occurred during the 2021-2022 cyclone season. This large set of
148 forecasts has then been analysed in order to assess the impact of the coupling in terms
149 of track, intensity and structure of the TCs.

150 The plan of the paper is as follows. Section 2 presents the forecast system AROME-
 151 IO, the experimental coupled system AROME-IO/NEMO and the description of the dif-
 152 ferent numerical simulations performed in this study. The results of these simulations
 153 are analysed in section 3. First, the statistical impact of the 3D OA coupling and the
 154 choice of initial conditions is assessed with scores against the Best-Track (BT) data for
 155 a large number of TC simulations. Second, the main conclusions of scores analysis are
 156 illustrated with three TC cases, Gelena (quasi-stationary) and Belna (regular propaga-
 157 tion velocity of about 5 m s^{-1}), and Batsirai, for which buoy measurements enable a di-
 158 rect comparison of the modelled cooling effects with observations. Conclusions and per-
 159 spectives are given in section 4.

160 **2 Tools, methods and data**

161 In this section, we give a short description of the numerical systems used for this
 162 study. We then describe the 5 different types of simulations which have been performed
 163 for 31 initial dates and which are analysed in section 3.

164 **2.1 The AROME-IO operational system**

165 AROME-IO (see Faure et al. (2020) and Bousquet et al. (2020) for more details)
 166 is an overseas version of the AROME-France model, which is the operational convection
 167 permitting, limited area, numerical weather prediction model used at Météo-France since
 168 2008. Since 2016, AROME-IO produces 48 hours forecasts (78 h on demand) 4 times a
 169 day over an area of 3000 km by 1400 km (30°E - 70°E , 7°S - 22°S) encompassing most is-
 170 lands of the SWIO. Its horizontal resolution is 2.5 km and the time step is 60 s. In the
 171 vertical, 90 levels are distributed between the surface (first level at 5 m and 34 levels within
 172 the first 2 km) and 10 hPa. The AROME-IO operational configuration is initialised from
 173 the High RESolution Integrated Forecasting System (HRES IFS) European Centre for
 174 Medium-Range Weather Forecasts (ECMWF) model (currently about 9km horizontal
 175 resolution), which also provides the lateral boundary conditions at the frequency of 1h.
 176 An IAU (Incremental Analysis Update; Bloom et al., 1996) initialisation scheme is used
 177 to built the small-scale features and to reduce the model spin-up at initial time t_0 . The
 178 IAU algorithm combines the ECMWF larger scale analysis increments (temperature, wind,
 179 humidity, and surface pressure) valid at t_0 with a 6 hour AROME-IO forecast initialised
 180 directly from the ECMWF analysis at $(t_0 - 6)$ hours.

181 The atmosphere-sea surface exchanges are represented by the Exchange Coefficient
182 Unified Multi-campaign Experiments (ECUME) parameterization (Weill et al., 2003; Be-
183 lamari, 2005) which is part of the EXternalised SURface platform SURFEX (Voltaire
184 et al., 2017). Moreover, AROME-IO is coupled every 300 s to a 1D OML model (Gaspar
185 et al., 1990; Lebeau-pin Brossier et al., 2009) in order to better take into account the feed-
186 back between the atmosphere and the ocean in cyclonic conditions. In the current con-
187 figuration, the OML prognostic parameters (salinity, temperature and currents) are ini-
188 tialised from the first 26 levels (depth of about 180 m) of the MERCATOR-Ocean global
189 product PSY4 (Lellouche et al., 2018) which are available every 6 hours (1h means).

190 The 1D OML parameterization is built on a partial representation of the ocean pro-
191 cesses which are triggered by winds. The advection and pressure gradient terms are ne-
192 glected in the system of dynamic and thermodynamic equations. The water columns are
193 therefore independent of each others. As a consequence, there is no vertical velocity gen-
194 erated by the divergence or convergence of currents and thus no upwelling and down-
195 welling. The OML model reproduces only the turbulent exchanges between the ocean
196 and the atmosphere and the turbulent mixing thanks to a 1.5 order turbulent scheme
197 directly adapted for the ocean from the Turbulent Kinetic Energy (TKE) scheme used
198 in AROME (Bougeault & Lacarrere, 1989).

199 In a limited area model, the TCs track are mostly driven by the large scale envi-
200 ronment of the TCs. In AROME, they usually follow the IFS scenario. The score of the
201 short term AROME forecast in term of TC position error is then very similar to the one
202 of the HRES IFS (Fig 1 (a)). The mesoscale model AROME significantly improves the
203 short term forecast of TC intensification compared to the HRES-IFS model which reg-
204 ularly underestimates TC intensities (Fig 1 (b-c)). AROME shows however a general ten-
205 dency to overestimate the TC intensification of the most intense systems. One of the ex-
206 pectation of the better representation of the interaction between the cyclonic circulation
207 and the oceanic surface in the OA coupled model is actually a reduction of the TC in-
208 tensity due to the generation of a cold upwelling underneath. We will see in the follow-
209 ing sections that it actually contributes to the reduction of the positive bias error in in-
210 tensity, even if other factors may also explain AROME overactivity.

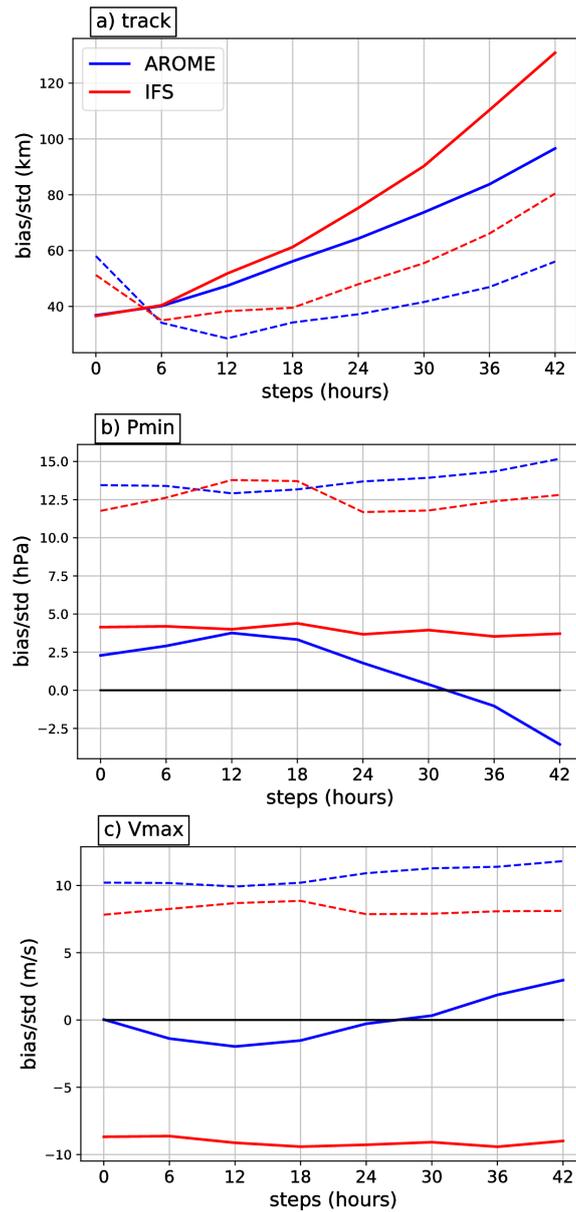


Figure 1.

Bias (solid lines) and standard deviation (dashed lines) for (a) the TC centre position, (b) the minimum of pressure at the centre of the TC and (c) maximal 10-m wind in the TC wall for IFS (red) and AROME-IO (blue) against the BT for all TCs of the 2018-2019 and 2019-2020 in the AROME-IO domain.

211 **2.2 The AROME-NEMO coupled system**

212 The experimental coupled system AROME-NEMO which has been implemented
 213 in the context of this work combines the AROME-IO atmospheric model (Faure et al.,

214 2020) and a regional version of the NEMO ocean model (Madec et al., 2019). The cou-
215 pling between these two components is controlled by the OASIS3-MCT coupler (Craig
216 et al., 2017).

217 NEMO is the European modelling framework for oceanographic research, opera-
218 tional oceanography, seasonal forecasting and climate studies. This study uses the ver-
219 sion 3.6 of the code with the same ORCA grid (tripolar grid with variable horizontal res-
220 olution; Madec & Imbard, 1996) at the horizontal resolution of a $1/12^\circ$ (about 9 km in
221 the SWIO region) and the same 50 unevenly spaced vertical levels as the PSY4 global
222 operational products of MERCATOR-Ocean. The NEMO domain covers the whole oceanic
223 part of AROME-IO. The bathymetry is based on the ETOPO1 database (Amante & Eakins,
224 2009). The vertical mixing is a TKE closure scheme based on the work of Gaspar et al.
225 (1990) (like the 1D OML used in this study), but with important modifications intro-
226 duced by Madec et al. (1998) in the implementation and formulation of the mixing length
227 scale.

228 NEMO provides to OASIS the 1h mean SST at a coupling frequency of 1 h. The
229 SST is used to compute the air-sea fluxes at each subsequent atmospheric time step. The
230 effect of surface currents on surface fluxes and atmospheric low-level flow is not consid-
231 ered in this study. Previous studies have shown little impact of currents on air-sea fluxes,
232 especially in comparison to the uncertainty associated with OA interactions in strong
233 winds (Pianezze et al., 2018; Bouin & Lebeaupin Brossier, 2020). The 1 h mean solar
234 and net heat fluxes and the components of the horizontal wind stress and atmospheric
235 freshwater are returned to OASIS by SURFEX and then sent to NEMO with a 1h cou-
236 pling frequency (Figure 2). The corresponding equations and the description of the cou-
237 pling strategy are detailed in Voltaire et al. (2017).

238 On the Météo-France supercomputers, the extra computing cost of the oceanic model
239 at a $1/12^\circ$ resolution and a coupling frequency of 1 hour is negligible compared to the
240 cost of the atmospheric model. For example, the current prototype runs on 480 proces-
241 sors for AROME-IO against only 32 processors for NEMO. So, the computing cost would
242 not be a limitation for a potential operational use of the coupled system. However, if this
243 configuration was meant to become operational with 4 runs a day, an important tech-
244 nical work would be needed in order to automate the processing of the NEMO initial and
245 lateral boundary condition files in real time. In the absence of a coupled data assimi-

246 lation system for AROME-IO, the design of the ocean initial condition in an operational
 247 suite with the coupled system is an open scientific question that we have started to ad-
 248 dress in the present work. But, the question of "coupled" initial conditions for the down-
 249 scaling of both an atmospheric (IFS) and oceanic (MERCATOR-Ocean) state is a com-
 250 plex subject in itself that will need to be further investigated prior to an operational im-
 251 plementation.

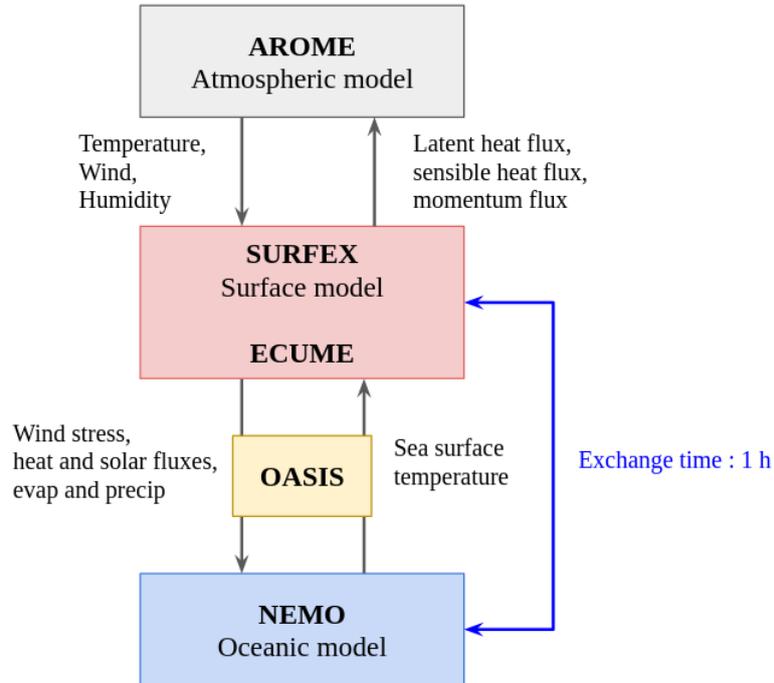


Figure 2. Schematic diagram of the coupling between AROME and NEMO via SURFEX and OASIS.

252 **2.3 Description of the TC simulations**

253 The main objective of this study is to evaluate the sensitivity of the AROME-IO
 254 TC forecasts skill to the representation of the ocean underneath.

255 Five different configurations have been implemented (Table 1). In the first config-
 256 uration, the SST field is initialised with the PSY4 ocean state and it is kept constant
 257 in the forecast. This configuration has been used in AROME-IO between 2016 and the
 258 end of 2017, prior to the implementation of the OML parameterization. It is also still
 259 the configuration of AROME-France.

Table 1. Summary of the 5 simulation types which are analysed in this work

	SST cst	OMLpsy4	OMLcyO	CPLpsy4	CPLcyO
Ocean configuration	None	1D OML	1D OML	3D ocean	3D ocean
Initial ocean state	PSY4	PSY4	Cycled Ocean	PSY4	Cycled Ocean

260 Then, the OML parameterization is switched on for two configurations and the new
 261 coupled system AROME-IO/NEMO is used for the last two configurations.

262 As an other objective of this work is to advice on the design of a future coupled
 263 operational system, two different solutions for the initial oceanic state have been used
 264 for both the OML and the coupled simulations. In a first solution, the ocean model di-
 265 rectly initialised with the PSY4 products provided by MERCATOR-Ocean. Such a so-
 266 lution is currently used in operation to initialise the OML at the beginning of each new
 267 forecast. In our study, we used this solution both for the OML configuration (later re-
 268 ferred to as OMLpsy4, which corresponds to the configuration of the current operational
 269 system) and the coupled AROME-NEMO configuration (referred to as CPLpsy4).

270 As the state of the ocean in the PSY4 global products is quite "smooth", the SST
 271 cooling and the upwelling generated by TCs is much weaker than in the AROME-NEMO
 272 coupled configuration (see discussion in section 3). Test simulations (not shown) with
 273 a forced configuration of NEMO using IFS or AROME winds showed that the main rea-
 274 son explaining the difference of SST cooling induced by TC circulations between PSY4
 275 and AROME-NEMO is the strength of the winds forcing the ocean. As pointed out by
 276 several recent studies, strong winds in the ECMWF IFS- WAVE Model (WAM) config-
 277 uration are generally underestimated (Pineau-Guillou et al., 2018; Haiden et al., 2018;
 278 Magnusson et al., 2019) whereas the wind from the current AROME-IO operational sys-
 279 tem are much closer to the wind estimated by the BT (Bousquet et al., 2020; see also
 280 Fig 1). We then decided to set up a second solution in which the ocean model is initial-
 281 ized by an ocean state resulting from a previous AROME-NEMO coupled run. In prac-
 282 tice, the coupled simulations start about a week before the TC enters the AROME do-
 283 main. The first coupled simulation is initialised with PSY4, then, 24 h later, the next

284 NEMO simulation starts from the state of the ocean of the 24 h forecast of the previ-
 285 ous coupled simulation (configuration referred to as CPLcyO). The "cycled" state of the
 286 ocean is also used to initialise the OML of an other series of simulations referred to as
 287 OMLcyO. Cycling the ocean without any relaxation toward observation is a solution which
 288 can be used only for a few days in a row. But, in the context of the TCs simulations con-
 289 ducted for this study, we think that this second solution emulates a forecast suite where
 290 the ocean would be initialised by an ocean analysis forced with the AROME winds in-
 291 stead of the IFS winds or by a combination between a smooth ocean analysis and an oceanic
 292 state issued from a previous coupled AROME-NEMO run.

293 In order to focus on the sensitivity to the ocean initial conditions, all five config-
 294 urations start from the same atmospheric state which is taken from the operational AROME-
 295 IO model. This means that the 6 h warmup is done with the OMLpsy4 configuration
 296 and the result is used as the initial condition of all five configurations. A few tests have
 297 been done with the same configuration both in the warmup and the 72h forecast. But,
 298 the spread in the TC characteristics after the warmup made the comparison between the
 299 different oceanic configuration difficult, hence our choice to start all configurations with
 300 the same atmospheric state for this study.

301 For each configuration, we have then set up a "light" NWP suite with only one 72 h
 302 forecast per day, at 00 UTC. The CPLcyO and OMLcyO suites start a few day before
 303 a TC enters the AROME-IO domain. The five suites have been run for seven different
 304 TCs that have been selected for various tracks and intensities (see Table 2) and with rea-
 305 sonable track prediction. The tracks of the seven cyclones are shown in Figure 3 and their
 306 intensity is indicated by colour dots corresponding to the five categories used in the SWIO
 307 basin: Tropical Depression (TD, $13 < V_{max} < 16 \text{ m s}^{-1}$), Moderate Tropical Storm (MTS,
 308 $17 < V_{max} < 24 \text{ m s}^{-1}$), Strong Tropical Storm (STS, $25 < V_{max} < 32 \text{ m s}^{-1}$), Trop-
 309 ical Cyclone (TC, $33 < V_{max} < 43 \text{ m s}^{-1}$), and Intense Tropical Cyclone (ITC, V_{max}
 310 $> 44 \text{ m s}^{-1}$) where V_{max} is the 10 min averaged maximum wind speed.

311 For each suite, 31 forecasts have been produced; grouping seven TCs, with one to
 312 six starting times each, corresponding to the 31 initial times given in Table 3. As in the
 313 current operational system, large initial intensity errors are sometimes inherited from
 314 the IFS analysis despite the 6 h warmup. In such cases, we focus on the differences be-
 315 tween the five configurations rather than on a comparison with the BT.

Table 2. Name and characteristics of the selected TCs.

	SWIO classification	Maximum average wind (m s^{-1})	US Saffir-Simpson scale
Gelena	Intense tropical cyclone	57	CAT3
Idai	Intense tropical cyclone	49	CAT2
Kenneth	Intense tropical cyclone	57	CAT3
Belna	Tropical cyclone	42	CAT1
Calvinia	Tropical cyclone	33	CAT1
Diane	Moderate tropical storm	21	Tropical storm
Herold	Intense tropical cyclone	46	CAT2
Batsirai	Intense tropical cyclone	57	CAT3

Table 3. List of the 31 forecasts used for the statistics. Dates in bold include a period when the TC is quasi-stationary (slow moving): translation speed $\leq 2 \text{ m s}^{-1}$ for at least 12 h.

Gelena	Idai	Kenneth	Belna	Calvinia	Diane	Herold
05/02/2019	09/03/2019	23/04/2019	06/12/2019	26/12/2019	23/01/2020	13/03/2020
06/02/2019	10/03/2019		07/12/2019	27/12/2019	24/01/2020	14/03/2020
07/02/2019	11/03/2019		08/12/2019	28/12/2019	25/01/2020	15/03/2020
08/02/2019	12/03/2019		09/12/19	29/12/2019		16/03/2020
09/02/2019	13/03/2019			30/12/2019		17/03/2020
10/02/2019	14/03/2019					18/03/2020

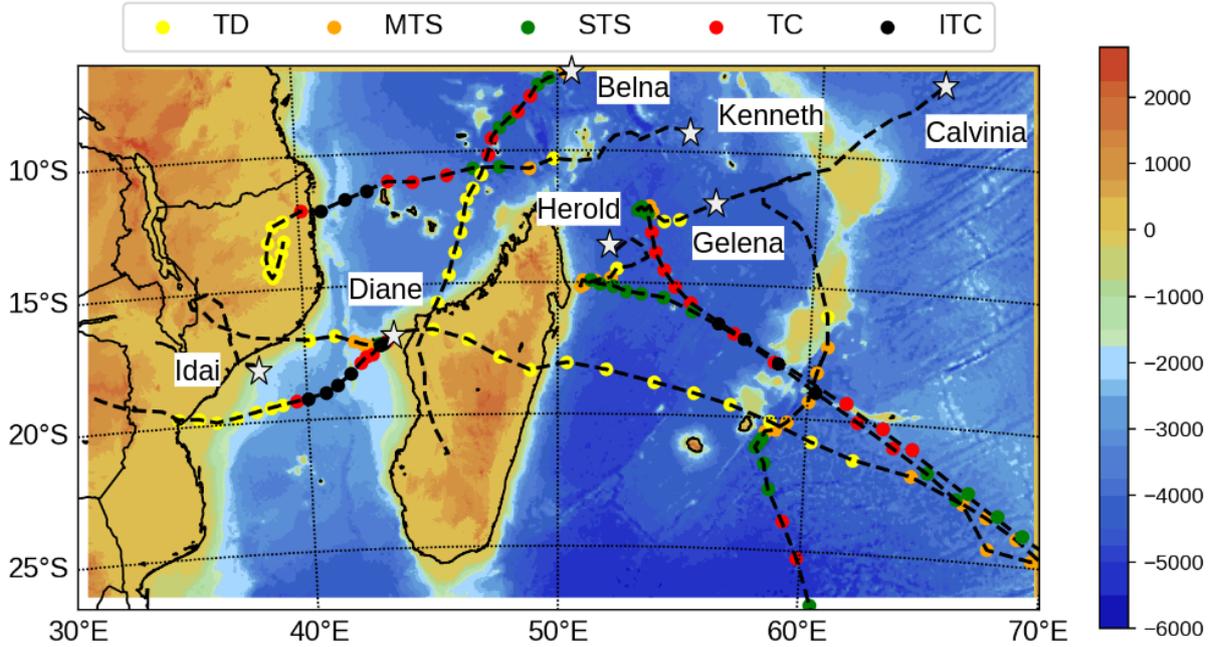


Figure 3. Orography and bathymetry (shading, in m) of the AROME-IO/NEMO coupled system. Track and intensity of Gelena, Idai, Kenneth, Belna, Calvinia, Diane and Herold as estimated by the RSMC of La Reunion. See text for the definition of the SWIO intensity classification.

316 **3 Results**

317 The aim of this section is to evaluate the coupled OA configuration and estimate
 318 its viability in an operational forecasting context. The first part of the section is a sta-
 319 tistical analysis of the 31 runs which have been produced for each of the 5 model con-
 320 figurations. The second part of the section focuses on two particular cyclone cases which
 321 have been chosen according to their translation speed, the oceanic conditions at the be-
 322 ginning of the simulation and their impact on the ocean.

323 Compared to the Atlantic or North Pacific basin, the SWIO basin is very poor in
 324 both atmospheric and oceanic observational data. We then mainly use the BT data for
 325 the evaluation of the track and intensity of the TC. When available, we will also add com-
 326 parison to the the SAR (Synthetic Aperture Radar) observations for the evaluation of
 327 stronger surface winds (available at <https://cyclobs.ifremer.fr/app/>; Mouche et al., 2017).
 328 Ocean sensors such as ARGO profilers or moored buoys (available at <http://www.coriolis.eu.org/Data-Products/Data-selection>) are rare and often far from the cyclonic area of interest. SST
 329

330 measurements from Earth observation satellites are usually very incomplete because of
331 the heavy cloud cover in TC condition. It is therefore quite difficult to provide an ac-
332 curate validation of the temperature predictions of ocean models in cyclonic conditions.
333 A comparison between the modelled oceanic state and observation from a drifting buoy
334 has been exceptionally possible in the case of TC Batsirai. It is shown in section 3.4.

335 **3.1 Scores against the Best-Track**

336 The TCs are tracked in all 31 forecasts of each of the 5 configurations (Table 3 and
337 Table 1).

338 The TCs minimum mean sea level pressure (Pmin) and Vmax are then compared
339 to the data of the objective analysis of the RSMC-La Réunion which is considered as the
340 reference in this study (it is later referred to as BT). The Pmin bias and standard de-
341 viation against the BT are shown in Figure 4. Conclusions are very similar for Vmax
342 (not shown).

343 Both constant SST runs (green curves) and OML runs (blue curves) overestimate
344 the intensity of the TCs with a bias of more than 10 hPa after 72 hours of simulation.
345 The OML simulations show however a slightly reduced bias between +48 and +60 hours
346 of simulation and a better standard deviation than the constant SST runs. As expected,
347 the systematic bias is significantly reduced by the 3D coupling with NEMO (pink curves).
348 The standard deviation after 30 hours of simulation is also improved in the 3D coupled
349 forecasts. The TCs characteristics in the OMLcyO simulations remains close to the ones
350 in the CPLcyO runs for about 24 hours but then, the scores of the OMLcyO runs quickly
351 degrade and reach the same bias and standard deviation as the runs with the operational
352 configuration OMLpsy4.

353 As we expect the slow or quasi-stationary TCs to be the most sensitive to the 3D
354 coupling with NEMO, we have split the original 31 dates into 2 groups. A first group,
355 later referred to as slow TCs, contains runs for which the translation speed of the TC
356 is less than or equal to 2 m s^{-1} for at least 12 hours (10 runs in Table 3). The remain-
357 ing runs (21 runs) form a second group of regular or fast moving TCs. For simplicity,
358 we will later referred to group 2 as "fast" TCs . The statistics have been recomputed
359 for each group (Figure 5).

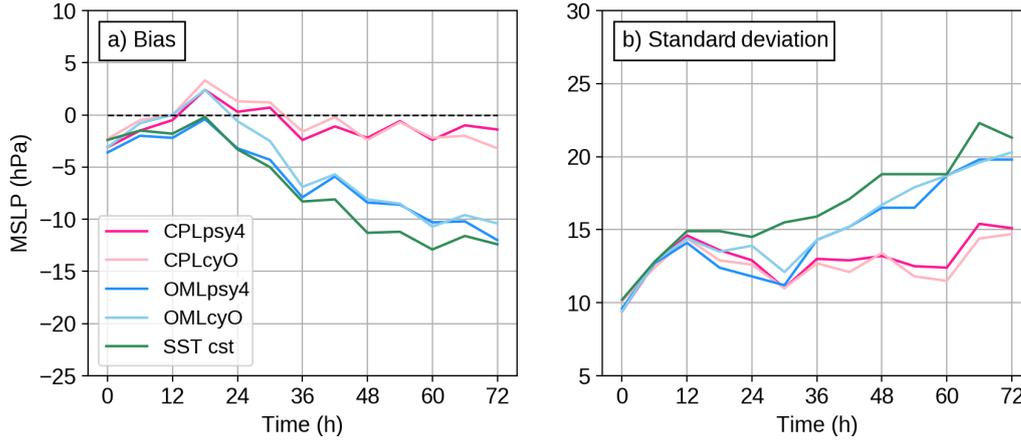


Figure 4. a) Bias and b) standard deviation for Pmin (hPa) derived from CPLpsy4 (dark pink), CPLcyO (light pink), OMLpsy4 (dark blue), OMLcyO (light blue) and SST constant (green) simulations against the BT. The statistics are based on 31 runs (corresponding to the different initial times) for 7 different TCs.

360 Figure 5 shows the Pmin bias (left) and standard deviation (right) against the BT
 361 data for the slow (top) and fast (bottom) groups. It clearly confirms that the 3D cou-
 362 pling between AROME and NEMO really matters for the intensity of the slow-moving
 363 storms with the bias of the OML simulations 20 hPa larger after 72 hours than the one
 364 in the CPL simulations. The impact of the 3D coupling is much smaller for the fast-moving
 365 TCs, with a tendency for the coupled runs, but also the OML run starting from the cy-
 366 cled ocean state to weakly underestimate the intensity of the TCs in the first 24 h. The
 367 behaviour of the model at the beginning of these simulations may be improved in a con-
 368 figuration where the 6 h IAU warmup uses the same configuration as the forecast. It is
 369 not the case here as all runs starts with the same initial atmospheric condition that is
 370 the result of the IAU warmup of the operational configuration (OMLpsy4). The coupled
 371 runs and OMLcyO may suffer from a spin-up period as the atmospheric boundary layer
 372 and the ocean adjust to each other and thus delay the intensification. More work would
 373 be needed to improved the consistency between the atmospheric and oceanic initial con-
 374 ditions in order to palliate the lack of OA coupled data assimilation, but this is beyond
 375 the scope of the present study.

376 In summary, the scores against the BT suggest that :

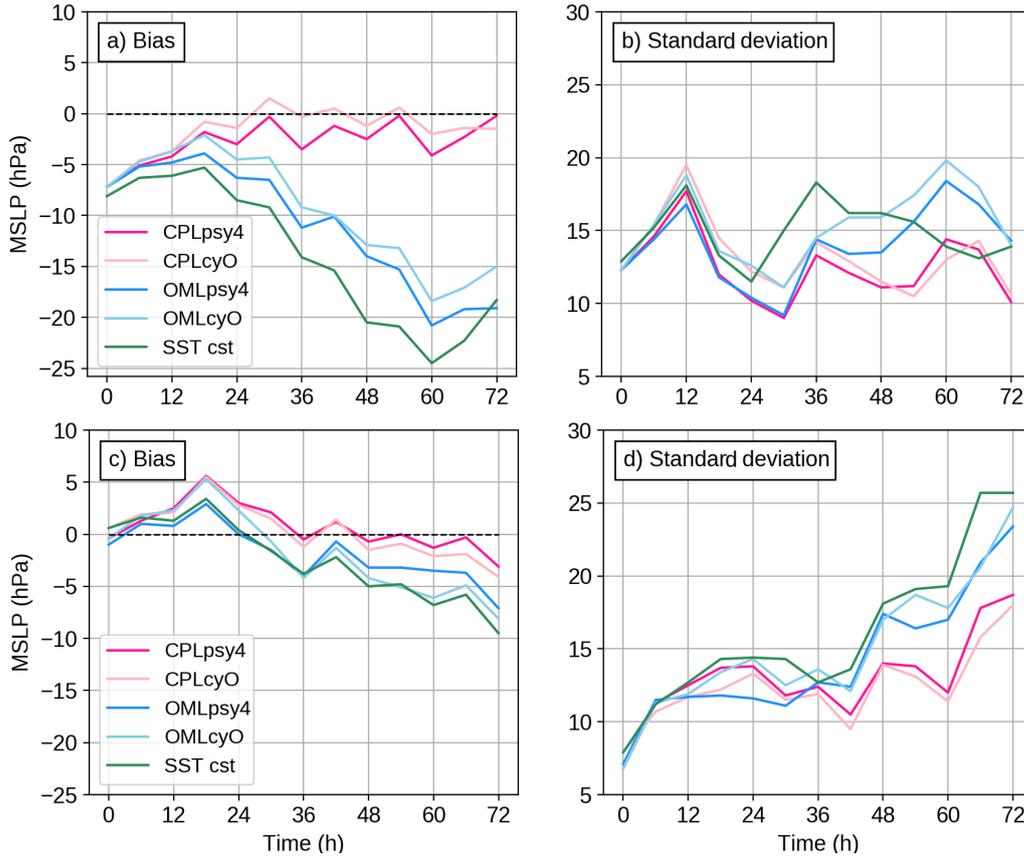


Figure 5. (a-c) bias and (b-d) standard deviation for Pmin (hPa) derived from CPLpsy4 (dark pink), CPLcyO (light pink), OMLpsy4 (dark blue), OMLcyO (light blue) and SST constant (green) simulations against the BT for (a-b) slow-moving storms (TC velocity $\leq 2 \text{ m s}^{-1}$ for a minimum of 12 h) and (c-d) fast-moving storms.

- 377 • The 3D coupling statistically improves the estimation of the TC intensification
- 378 in AROME-IO after 24/30 h of simulation,
- 379 • but the OA feedback allowed by the 3D coupling only significantly impact the in-
- 380 tensity of slow storms.
- 381 • The cooling in the oceanic initial conditions that is consistent with the AROME
- 382 wind forcing only impacts the first 24 hours of the TC intensity forecast when the
- 383 OML parameterization is later used in the forecast model (OMLcyO).

384 In the following section, we analyse with more details the oceanic and the atmo-

385 spheric responses in the new OA coupled system for three particular cases: TC Gelena

386 from the "slow" group, TC Belna from the "fast" group and TC Batsirai for which we
387 have drifter observations.

388 **3.2 Gelena: a case of extreme cooling of the ocean in AROME-NEMO**

389 **3.2.1 Gelena analysis by the RMSC-La Réunion**

390 Gelena was the ninth tropical storm of the 2018-2019 cyclone season in the SWIO.
391 Gelena formed within an active monsoon trough taking place in early February 2019 over
392 the SWIO from an area of low pressure located less than 1000 km east of Diego-Suarez
393 (northern tip of Madagascar). It encountered a rapid development late on 5 February
394 2019 and the system became a TS in the early hours of 6 February. It reached the level
395 of STS, almost TC (equivalent CAT2 on the US Saffir-Simpson scale) at the end of 6 Febru-
396 ary. It weakened on 7 February due to the cooling of the SST induced by its very slow
397 motion on 6 and 7 February. Gelena accelerated after 7 February and resumed its in-
398 tensification. It became a TC on the morning of 8 February, then it reached the stage
399 of ITC (equivalent CAT3) 24 hours later. Gelena reached its maximum intensity in the
400 early afternoon of 9 February with maximum winds estimated over 10 minutes at 57 m
401 s^{-1} (equivalent CAT4) and a minimum sea level pressure of 938 hPa before weakening
402 due to vertical shear. At the end of the night of 9 February, Gelena approached within
403 60 km of Rodrigues Island as ITC. The island suffered violent wind gusts recorded at
404 46 m s^{-1} at Pointe-Canon. During the following days, Gelena moved along a south-east
405 and then east-south-east track and slowly filled in and dissipated at the end of 16 Febru-
406 ary east of 90°E in the subtropics.

407 Both the operational models IFS and AROME-IO overestimated the intensity of
408 TC Gelena compared to that estimated by the BT. We suspected that the quasi-stationary
409 behaviour of Gelena on 6 February and most part of 7 February led to a strong oceanic
410 response which could have been underestimated by the IFS and the current OMLpsy4
411 configuration of AROME-IO. We will see in the next section that AROME-NEMO ac-
412 tually triggers a very intense upwelling on 6 February. After 24 h of simulation, the state
413 of the ocean in the 6 February run is much cooler than the corresponding PSY4 anal-
414 ysis. The run of 7 February 2019 is then a good case to analyse the impact of the oceanic
415 initial condition as the cyO and psy4 simulations do start with significantly different ocean
416 conditions.

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3.2.2 Simulations of TC Gelena

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In this section, the development of TC Gelena and the corresponding response of the ocean are analysed with 2 sets of 5 runs starting 6 February 2019, 00 UTC and 7 February 2019, 00 UTC.

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The TC tracks (Figure 6a) simulated with the CPL (pink curves), OML (blue curves) and SST constant (green line) configurations are similar and close to the BT that is estimated by the RSMC-La Réunion (black line), despite a slight shift to the west after the 8 February 2019, 06 UTC. The mean error (including both along track and across track errors) is approximately 30 km during the first 30 h of the simulations and it reaches about 200 km at the end of the 72h simulations .

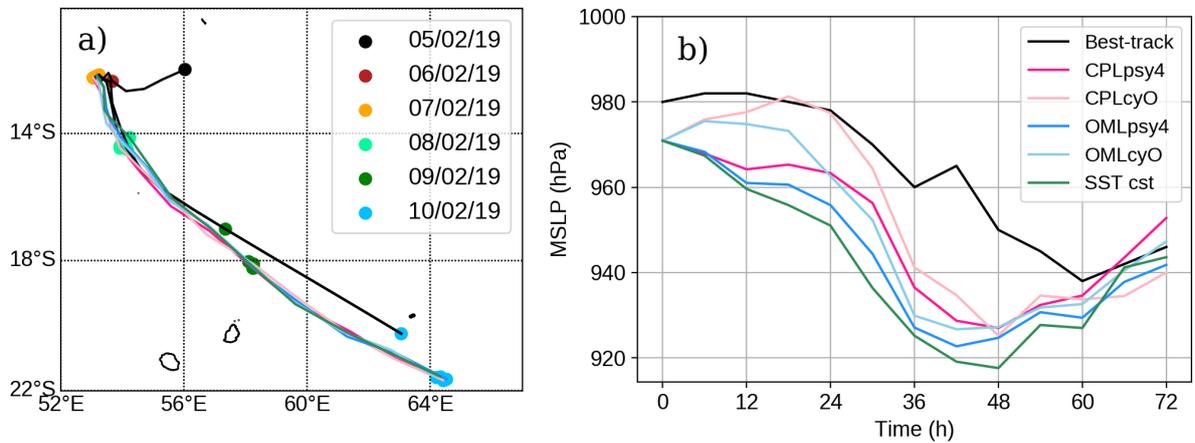


Figure 6. (a) Track of Gelena and (b) time evolution of Pmin (hPa) from the 7 February 2019 at 00 UTC till the 10 February 2019 at 00 UTC (72 h) for the RSMC La Réunion BT data (black) and the CPLpsy4 (dark pink), CPLcyO (light pink), OMLpsy4 (dark blue), OMLcyO (light blue) and SST constant (green) configurations.

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Gelena has a quasi-stationary behaviour on the 6 February 2019 with an average translation speed of 1.5 m s^{-1} . On the 7 February 2019, Gelena maintains a slow translation speed of 3 m s^{-1} and then it accelerates during the next 48 h, with an average translation speeds of 8 m s^{-1} .

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In the 3D coupled runs of the 6 February 2019 at 00 UTC, the cooling of the SST reaches about 6°C in the first 24 hours of the simulations. Such an intense cooling is not present in the MERCATOR product PSY4 of the 7 February 2019 at 00 UTC. The ini-

434 tial state of the ocean for the psy4 runs starting on the 7 February 2019 at 00 UTC is
435 then significantly warmer than the one of the cyO runs (Figure 8(a-c)) and 9(a-c)).

436 Figure 6(b) shows the temporal evolution of Pmin (hPa) derived from the RSMC
437 La Reunion BT data (black), the CPLpsy4 (dark pink), CPLcyO (light pink), OMLpsy4
438 (dark blue), OMLcyO (lightblue) and SST constant (green) simulations, for the runs start-
439 ing on the 7 February 2019 at 00 UTC. The impact of cycling the ocean state on the TC
440 intensity is clearly seen on the first 24 hours of the simulations as the psy4 runs are about
441 10 hPa deeper than the cyO runs. In this case, all runs highly overestimates the inten-
442 sification after 30 h of simulation with a simulated maximum of intensity after about 36
443 to 48 h of simulation. At the peak of intensity, Pmin in the CPL runs is however about
444 5 hPa weaker than in the OML runs. The simulation with constant SST (green) shows
445 that neglecting completely the feedback between the TC and the ocean generates in this
446 case an even larger overestimation of maximum TC intensity of about 5 hPa compared
447 to the OMLpsy4 simulation.

448 Figure 7 shows the surface (10 m) wind speeds of TC Gelena in the run starting
449 on 7 February 2019 for the CPLcyO and OMLpsy4 simulations compared to the SAR
450 data. The surface wind field observed by the SAR shows a maximum wind speed value
451 in the eyewall of 37 m s^{-1} on 7 February 2019 at 02 UTC and about 35 m s^{-1} on the
452 8 February 2019 at 02 UTC. The winds simulated by the AROME model are much stronger
453 than those observed by the SAR data, especially after 26 hours of simulation with OMLpsy4
454 when the wind speed values in the eyewall reach about $55\text{-}60 \text{ m s}^{-1}$. The maximum wind
455 in CPLcyO is about 15 m s^{-1} weaker, but still more than 10 m s^{-1} higher than the ob-
456 servation.

457 The response of the ocean to the cyclonic winds in CPLpsy4, CPLcyO, OMLpsy4
458 and OMLcyO simulations starting on the 7 February 2019 is illustrated in Figures 8 to
459 11. In both CPL simulations (Figure 8(b-d)), the SST cools by another 6°C in the first
460 24 h of the simulation when the TC is still slow. In OMLpsy4, the SST cools only by
461 about 1.5°C (Figure 9b). In the OMLcyO simulation, the SST which had been cooled
462 by 6°C in the previous coupled run now heats up by 2°C (Figure 9d). We will see later
463 in this section that actually, in the first 12 h of simulation starting on the 7 February
464 2019, the OA surface fluxes in the CPLcyO and OMLcyO runs tend to heat up the ocean.

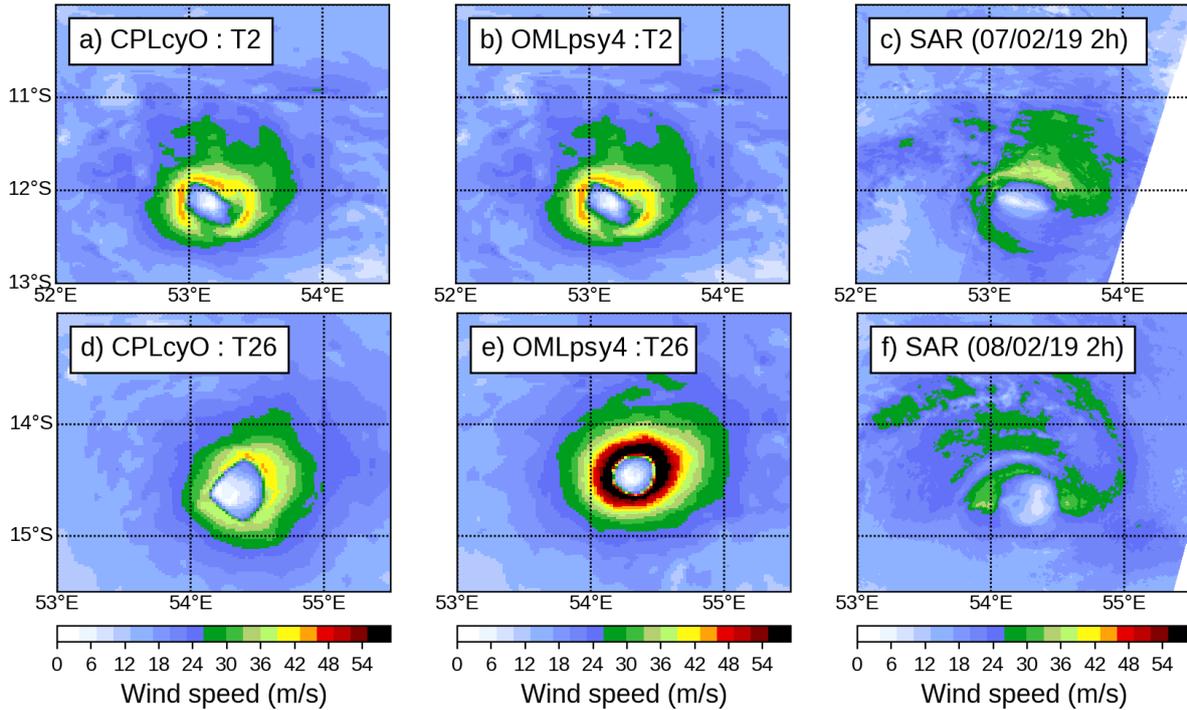


Figure 7. TC Gelena surface winds: (a-b) (resp. d-e) 10 m wind speed (m s^{-1}) of the 2 h (resp. 26 h) forecasts verifying on 7 February 2019, 02 UTC (resp. 8 February 2019, 02 UTC) for both CPLcyO and OMLpsy4; c) (resp. f) SAR surface wind speed (m s^{-1}) on 7 February 2019 at 02 UTC (resp. 8 February 2019, 02 UTC).

465 This difference in surface water cooling between the CPL and OML simulations is
 466 explained by the presence of an intense upwelling in the CPL runs (Figure 10). In CPLcyO,
 467 the upwelling is already intense in the initial condition. It has already crossed the bot-
 468 tom of the OML in the first 24 h of the previous forecast (CPLcyO from the 6 Febru-
 469 ary 2019 at 00 UTC) and it remains strong in the first 24 h of the 7 February 2019 00 UTC
 470 simulation cooling the surface by an other 6°C . In the PSY4 initial condition of CPLpsy4,
 471 the upwelling is broader and weaker and still confined to 50 m below the surface At the
 472 end of the quasi-stationary period on 8 February 2019 00 UTC, the surface of the ocean
 473 is 6°C cooler in CPLcyO than in CPLpsy4.

474 As expected in the South hemisphere, the upwelling in the CPL runs is located on
 475 the left (East) side of the TC track. In this zone, the thermocline vanishes and the warm
 476 water of the OML is completely replaced by deeper cold water. A zone of maximum warm-
 477 ing is observed on the East side of the cooling at a depth between 40 m and 80 m. It cor-

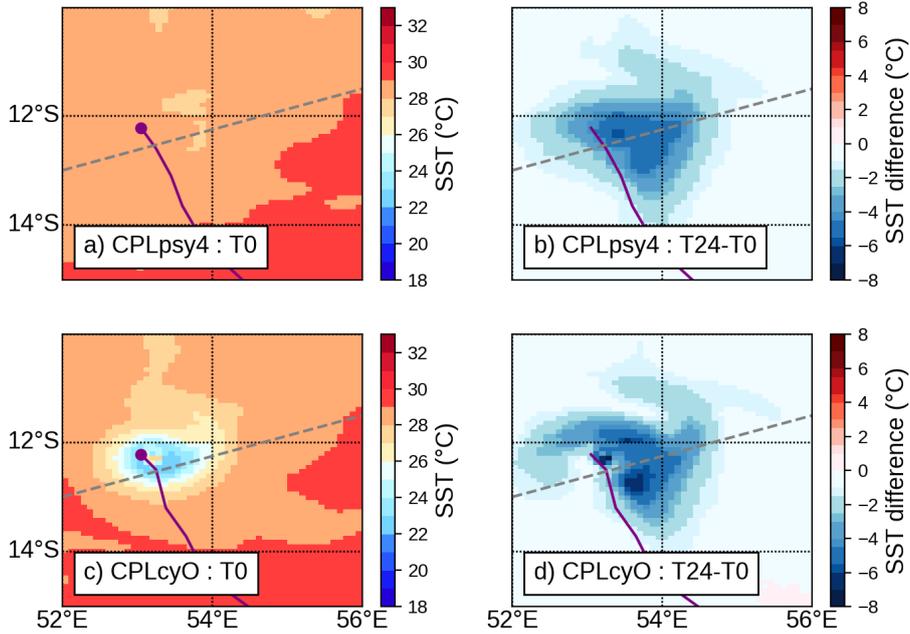


Figure 8. SST ($^{\circ}\text{C}$) in the vicinity of Gelena for the simulation starting on the 7 February 2019 at 00 UTC : (a-c) initial condition of simulation with the CPLpsy4 (top) and CPLcyO (bottom) configurations and (b-d) SST difference ($^{\circ}\text{C}$) between the initial condition and the 24 h forecast. The purple point gives the position of the cyclone at the time of the figure and the track of the cyclone is represented by the purple line.

478 responds to a zone where the turbulent mixing deepens the OML in this region as colder
 479 water originally under the initial thermocline is mixed with warmer water from the OML.

480 For the OMLpsy4 simulations (Figure 11), the cooling of the surface waters is very
 481 shallow, between the surface and a depth of about 20 m and it does not exceed 2°C . In
 482 OMLcyO, the intense upwelling that started in the first 24 h of the previous run can-
 483 not continue in an uncoupled run. It is slowly eroded from the top by the surface fluxes
 484 resulting in a weak warming of 1°C in the upper part of the upwelling which was present
 485 in the initial condition.

486 Figure 12 shows the surface latent heat fluxes (W m^{-2}) and the radial and tan-
 487 gential winds (m s^{-1}) at 50 m above the surface around the Gelena centre after 6 h of
 488 the simulation starting on the 7 February 2019 at 00 UTC for CPLcyO, CPLpsy4, OML-
 489 cyO and OMLpsy4. Both simulations that start with a cycled ocean show a reversal of

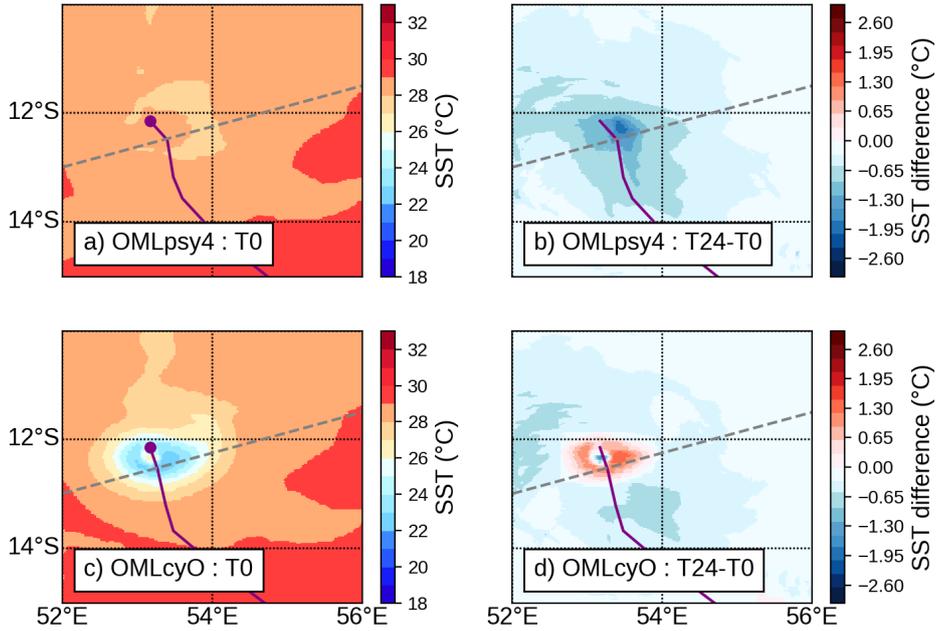


Figure 9. As Figure 8 but for OMLpsy4 and OMLcyO configurations. Note that the colour scale for the differences is different for fig. 9 and fig. 8

490 the heat fluxes at the OA interface of -250 to -500 W m^{-2} meaning that, when the ocean
 491 cooling is intense and fast, the atmosphere gives back heat to the ocean contrary to the
 492 primary mechanism leading to TC intensification. Such an inversion of sensible and la-
 493 tent heat fluxes between the air and the sea surface is also mentioned by Glenn et al. (2016).
 494 The minimum heat flux is observed in the left rear part of the TC track in the region
 495 of the strongest upwelling (as shown in Figure 8 and Figure 10). In OMLcyO, the in-
 496 version of heat fluxes sign under the TC (Figure 12d) causes a warming of the surface
 497 waters which is not compensated by the cooling associated with the turbulent mixing
 498 in the OML (Figures 9 and 11).

499 The processes below and at the surface discussed above have a direct impact on
 500 the vertical structure of the TC. Figure 13 illustrates the vertical structure of the mean
 501 state of Gelaya in its NW and NE quadrants, for the CPLcyO and OMLpsy4 simulations.
 502 These two quadrants show the rear area of the cyclone where the cooling was the most
 503 intense. The strong low level convergence simulated by OMLpsy4 results in a strong con-

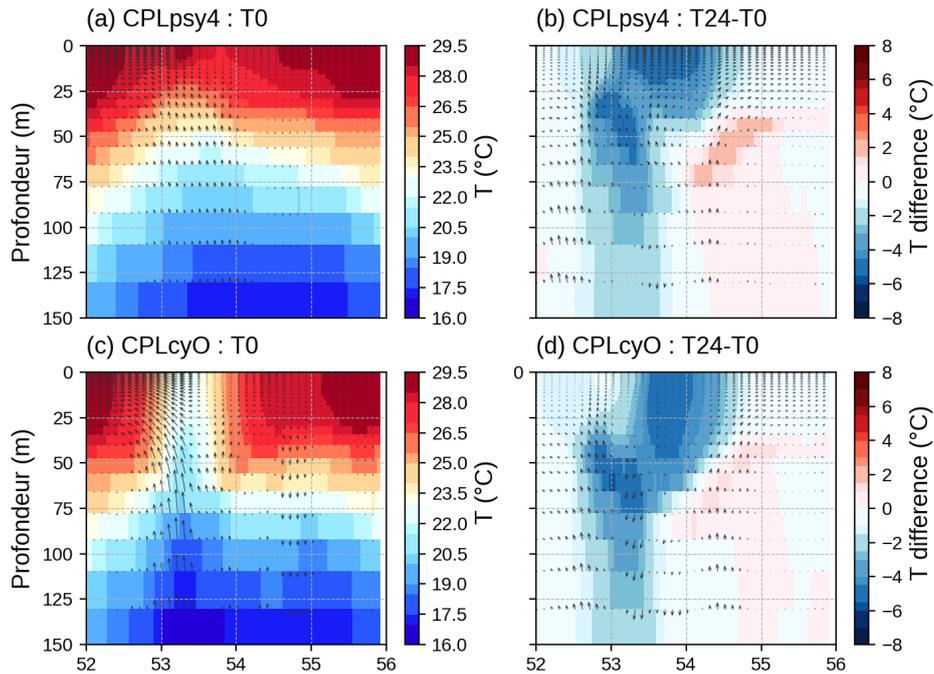


Figure 10. West-East cross-section along the dashed grey line on Figure 8 showing the evolution of the ocean temperature (T) in the run starting on the 7 February 2019 at 00 UTC run (a-c) initial condition of simulation with the CPLpsy4 (top) and CPLcyO (bottom) configurations and (b-d) temperature (T) difference ($^{\circ}\text{C}$) between the initial condition and the 24 h forecast

504 vective zone, which extends up to 16 km altitude in both quadrants. In this case, the
 505 deep convection shows a circular symmetry around the eye. The maximum of synthetic
 506 radar reflectivity that is associated with the eyewall extends up to 5 km above the sur-
 507 face. In CPLcyO, the convergent secondary circulation at the surface is weaker. The depth
 508 of the deep convection is reduced to 14 km height. In this case, the cyclone is more asym-
 509 metric with weaker convection in the NE part above the coldest surface waters. Low level
 510 clouds forms in the northern part of the eye where the heat fluxes are negative (Figure 13)
 511 while in OMLpsy4, the TC eye remains very dry. Low level clouds in the eye are a well
 512 known feature (see for example, Kossin et al., 2002 and Houze, 2010), but to our knowl-
 513 edge, their formation has not yet been linked to the ocean cooling under a TC.

514 In summary for the case of Gelena, all 5 model configurations overestimate the TC
 515 intensity. In the operational configuration OMLpsy4, the maximum wind near the max

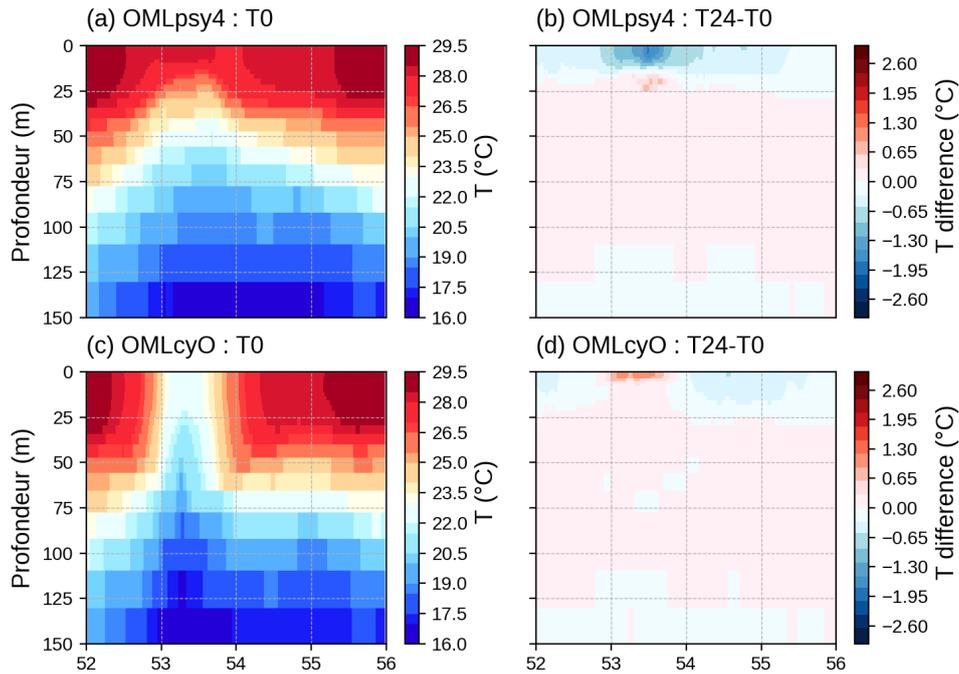


Figure 11. As Figure 10 but for OMLpsy4 and OMLcyO configurations. Note that the colour scale for the differences is different for fig. 11 and fig. 10.

516 of intensity is about twice the observed wind speed. The overestimation is reduced by
 517 a factor of two in the CPLcyO configuration. The feedback between the upwelling and
 518 the TC which is present only in the coupled configurations is a key factor in this case
 519 where the TC remains quasi-stationary for almost 48 h. But, if the coupling with a 3D
 520 ocean model partially solves the overestimation of the winds, other components of the
 521 model are probably still needed to improved the behaviour of the model in the case of
 522 intense TC (surface flux parameterization, coupling with waves/sea sprays, currents, cloud
 523 microphysics). This case also shows the importance of the choice of oceanic initial con-
 524 dition in short term forecasts. The memory of the oceanic circulations triggered by the
 525 high resolution TC winds from AROME is lost if the ocean in the coupled model is ini-
 526 tialised by the PSY4 products. The impact on the TC intensity is found to be signifi-
 527 cant in case of intense cooling.

528 In the case of Gelena, we could not find any in situ measurements to validate the
 529 SST cooling. It is then impossible to conclude on the realism of the 12°C cooling in 48
 530 hours produced by the CPLcyO configuration. References to such large cooling are found

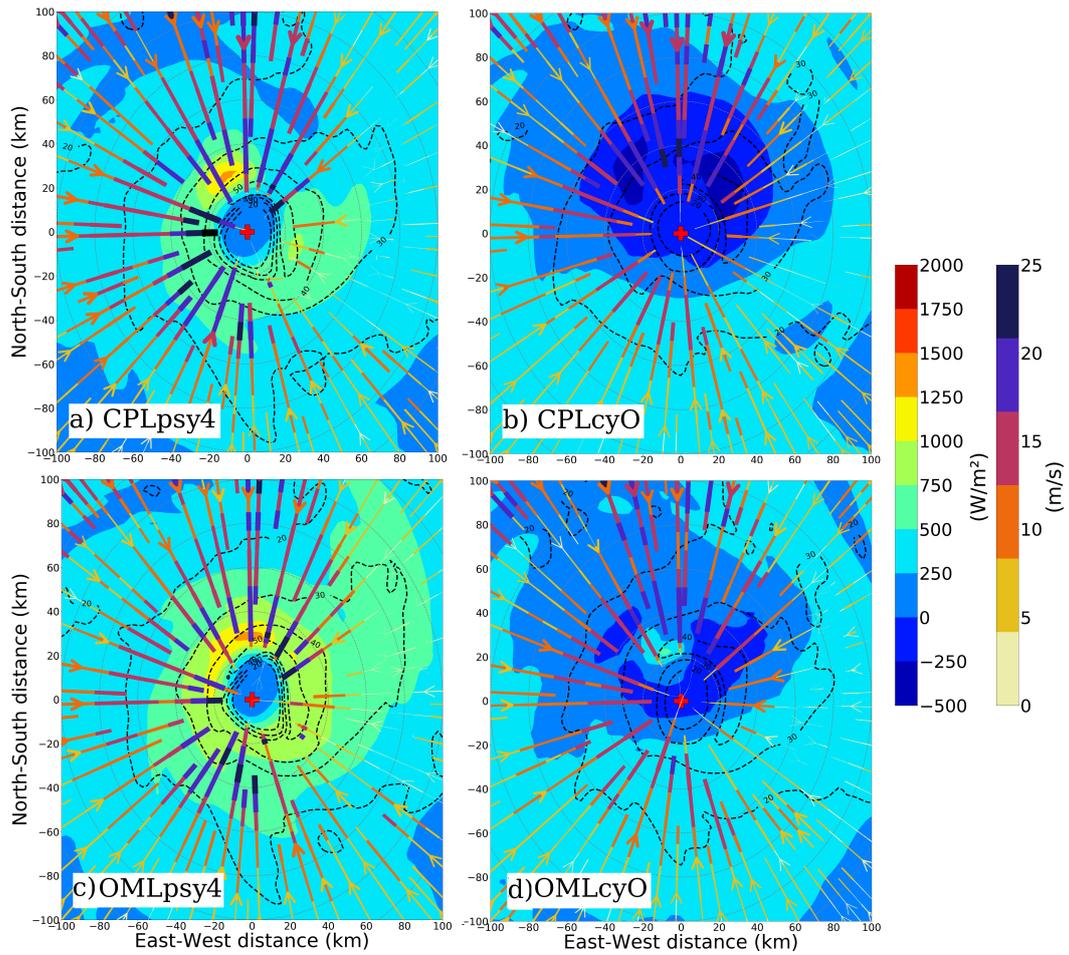


Figure 12. Surface latent heat fluxes (W m^{-2} , shading) near the centre of Gelena after 6 h of simulation of runs starting on the 7 February 2019 at 00 UTC : a) CPLpsy4, b) CPLcyO, c) OMLpsy4 and d) OMLcyO. The radial wind at 50 m above the surface (m s^{-1}) is represented by the coloured arrows and isolines of tangential wind speed (m s^{-1}) are drawn with black dashed contours.

531 in the literature (Chiang et al., 2011; Guan et al., 2021). Our belief is however that the
 532 ocean cooling that is obtained by the CPLcyO configuration in the case of Gelena is a
 533 bit overestimated as a consequence of the overintensification of the TC intensity in AROME-
 534 NEMO.

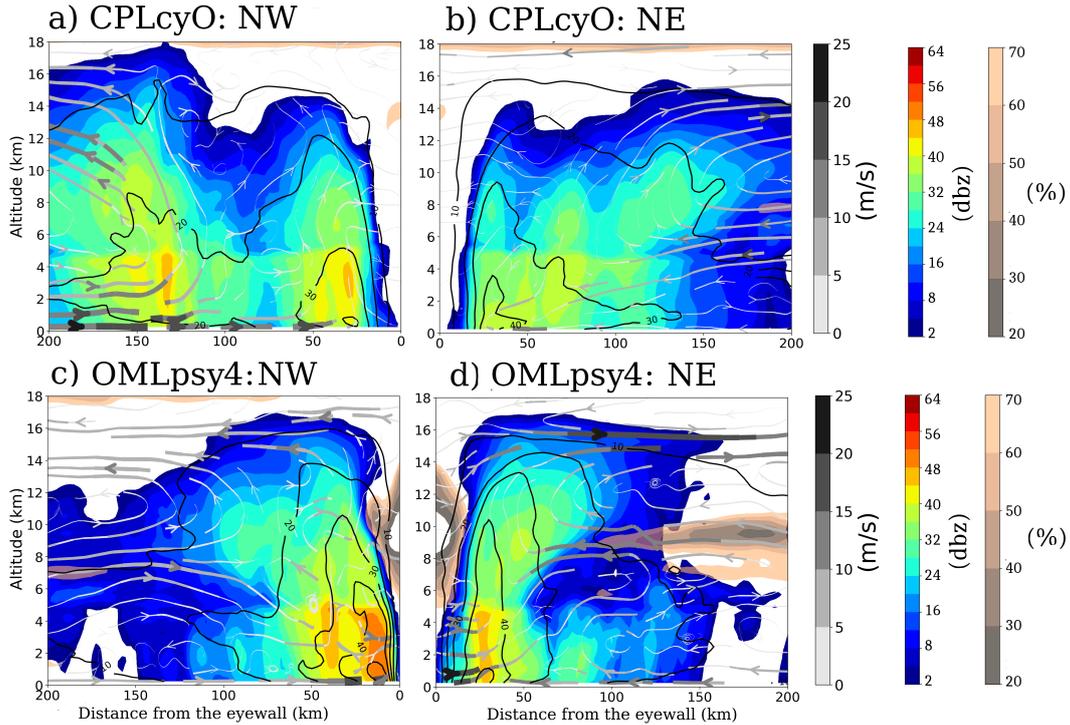


Figure 13. Vertical sections of the azimuthal mean of synthetic radar reflectivity (dBz) for the (a-c) North-West and (b-d) North-East quadrants of Gelena after 12 h of simulation from the 7 February 2019 at 00UTC with the OMLpsy4 and CPLcyO configurations. On the same figure, the secondary circulation is shown with streamlines in black and white and the tangential wind speed (m s^{-1}) with black isocontours. Relative humidity (%) is in pink/grey shadings.

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3.3 Belna: a case of steadily moving TC

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3.3.1 Belna analysis by the RMSC-La Réunion

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Belna developed within a low level convergence zone resulting from the conjunction of a Madden Julian oscillation phase and the passage of an equatorial Rossby wave in early December 2019 South of the Seychelles. During the night of 05 December, it developed from a moderate to a STS and became a TC on the morning of 07 December. Belna reached its maximum intensity in the late afternoon of 7 December with maximum estimated winds of 43 m s^{-1} and a minimum sea level pressure of 954 hPa. It maintained a south-western track and passed between the islands of Aldabra and Astove and 100 km west of Mayotte during the day on 08 December. It weakened to a STS during the first six hours of 09 December, and then re-intensified to a TC during the next six hours

546 north of Madagascar. During the following days, Belna weakened and it completely dis-
 547 sipated on 11 December.

548 TC Belna was not moving very fast, but it has a regular pace at about 4 m s^{-1} .
 549 In enters the AROME domain during the night of 5 December. We will focus on the run
 550 starting on the 6 February 2019 at 00 UTC. In this case, the initial oceanic state from
 551 PSY4 and from the cycled ocean configurations are very similar.

552 **3.3.2 Simulations of TC Belna**

553 In this section, the development of TC Belna and the corresponding response of
 554 the ocean are analysed with a set of 5 runs starting on the 6 December 2019 at 00 UTC.

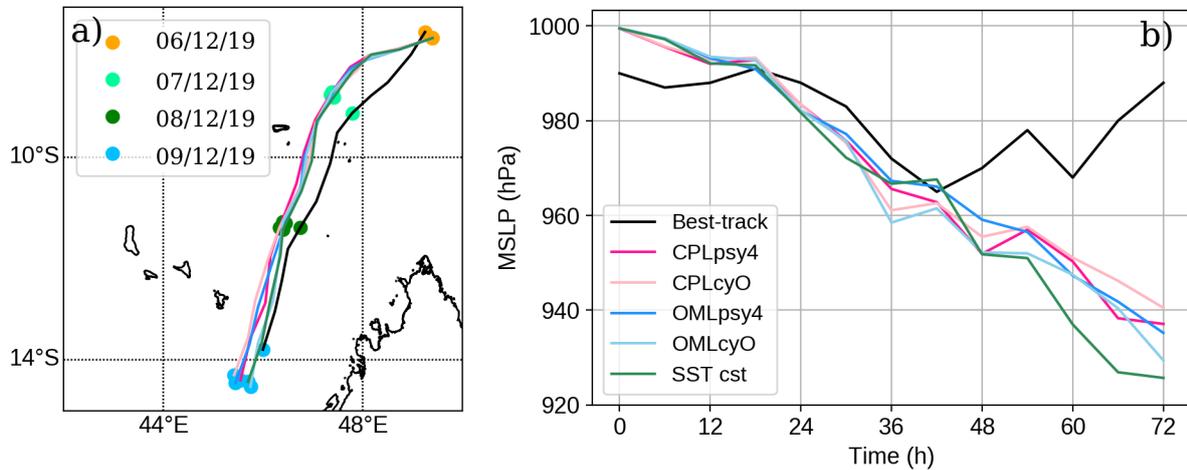


Figure 14. (a) Track of Belna and (b) Time evolution of Pmin (hPa) from the 06 December 2019 at 00 UTC to the 9 December 2019 at 00 UTC (72 h) for the RSMC La Réunion BT data (black) and the CPLpsy4 (dark pink), CPLcyO (light pink), OMLpsy4 (dark blue), OMLcyO (light blue) and SST constant (green) configurations.

555 Figure 14 shows the track of Belna and the time evolution of Pmin (hPa) as given
 556 by the RSMC La Reunion BT data (black) and derived from the CPLpsy4 (dark pink),
 557 CPLcyO (light pink), OMLpsy4 (dark blue), OMLcyO (lightblue) and SST constant (green)
 558 simulations. As in the case of TC Gelena, Belna tracks are very similar in the 5 simu-
 559 lations (Figure 14a) and they are close to the BT (black line), with a small westward er-
 560 ror in TC position in the first 48 h of the simulations. The average error is about 60 km

561 over the entire period. The simulated translation speed of Belna is on average 4 m s^{-1}
 562 during the whole simulation (not shown). During the first 36 hours, the Pmin forecast
 563 in the 5 configurations remains close to the BT analysis. But after 36 h, the TC contin-
 564 ues to intensify in all model configurations while Belna starts to weaken in the obser-
 565 vations.

566 The anomaly in the forecast is confirmed by the comparison with two SAR images
 567 (Figure 15). After 26 hours of simulation, the wind speed values in the eyewall are very
 568 similar to those observed on the SAR images, i.e. between 30 and 40 m s^{-1} . However,
 569 after 50 h of simulation, a few hours after the intensity peak, the wind speeds in the eye-
 570 wall are about 30 m s^{-1} higher in the CPLcyO and OMLpsy4 simulations than in the
 571 SAR data.

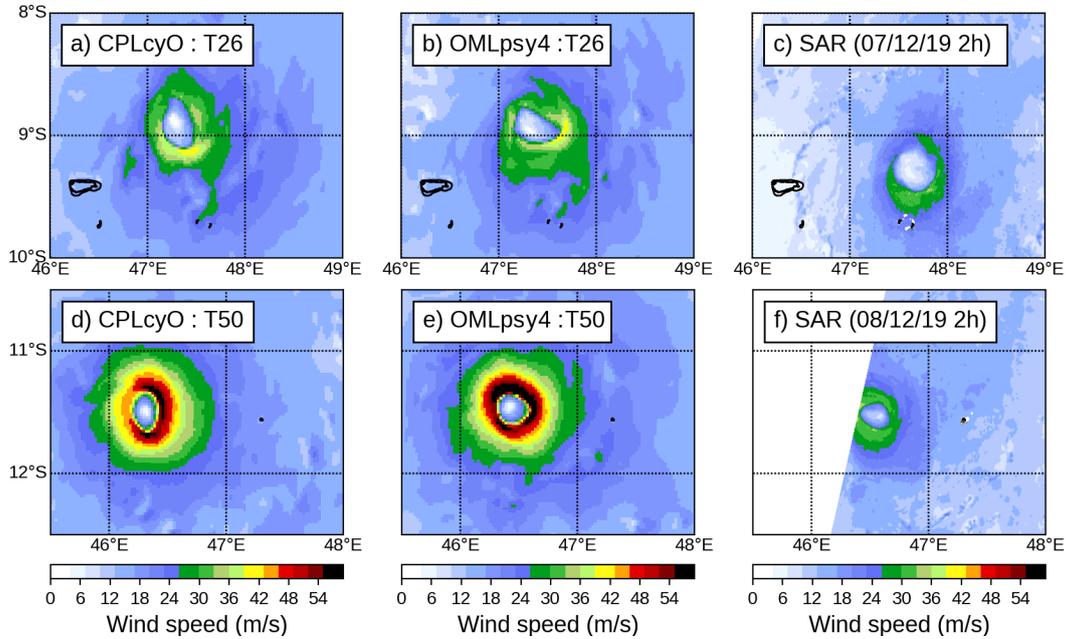


Figure 15. TC Belna surface winds: (a-b) (resp. d-e) 10 m wind speed (m s^{-1}) of the 26 h (resp. 50 h) forecasts verifying on 7 December 2019, 2 UTC (resp. 8 December 2019, 2 UTC) for both CPLcyO and OMLpsy4; c) (resp. f) SAR surface wind speed (m s^{-1}) on 7 December 2019 at 2 UTC (resp. 8 December 2019, 2 UTC).

572 The response of the ocean in CPLcyO and OMLpsy4 simulations starting on the
 573 6 December 2019 is illustrated in Figure 16 and figure 17. CPLpsy4 (resp. OMLcyO)
 574 is not shown because the results are very similar with CPLcyO (resp. OMLpsy4). The

575 CPLcyO and OMLpsy4 simulations start with very similar oceanic states (Figure 16a
 576 and c). After 48 hours of simulation, in both CPLcyO and OMLpsy4 the area of SST
 577 cooling around the TC is quite large, on both side of the track. The maxima of cooling
 578 are however still found on the left (East) side of the track, with a maximum value of 2°C
 579 in CPLcyO and OMLpsy4. The vertical cross-section shown in Figure 17 show a nar-
 580 row upwelling in the CPL run which weaken as it reaches the bottom of the OML. Most
 581 of the cold water remains confined under the OML at around 50 m depth. A weak ex-
 582 tension of upwelling in the OML may however explain the small difference in SST cool-
 583 ing between the CPLcyO and the OMLpsy4 runs. In this case, the upwelling does not
 584 destroy the thermocline, but the depth of the thermocline gains about 20 m in the vicin-
 585 ity of the TC as indicated by the positive temperature tendencies between 50 and 70 m
 586 in Figure 17(b-d).

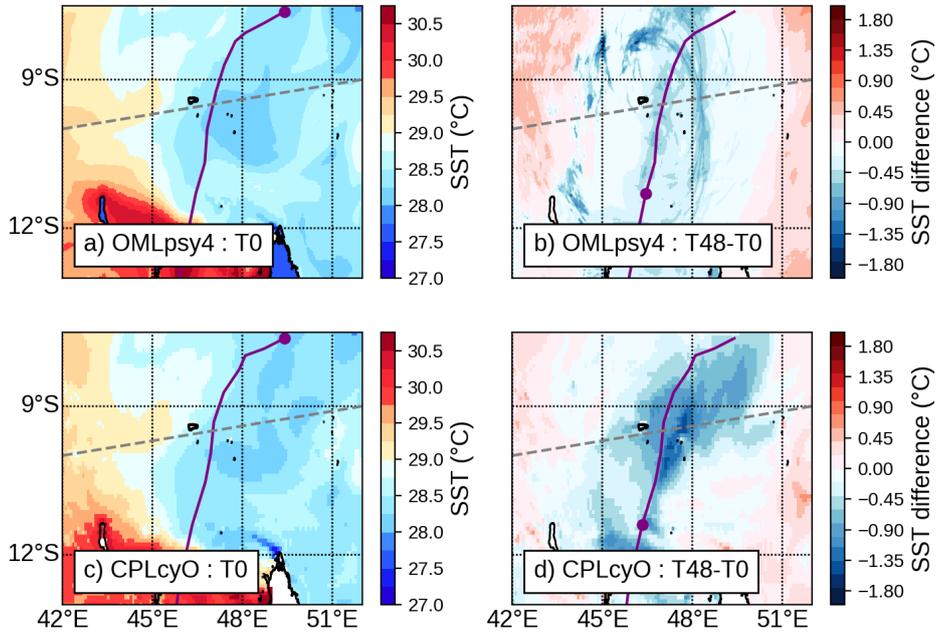


Figure 16. SST ($^{\circ}\text{C}$) in the vicinity of Belna for the simulation starting at 00 UTC on the 6 December 2019 : (a-c) initial condition of simulation with the OMLpsy4 (top) and CPLcyO (bottom) configurations and (b-d) SST difference ($^{\circ}\text{C}$) between the initial condition and the 48 h forecast. The purple point gives the position of the cyclone at the time of the figure and the track of the cyclone is represented by the purple line.

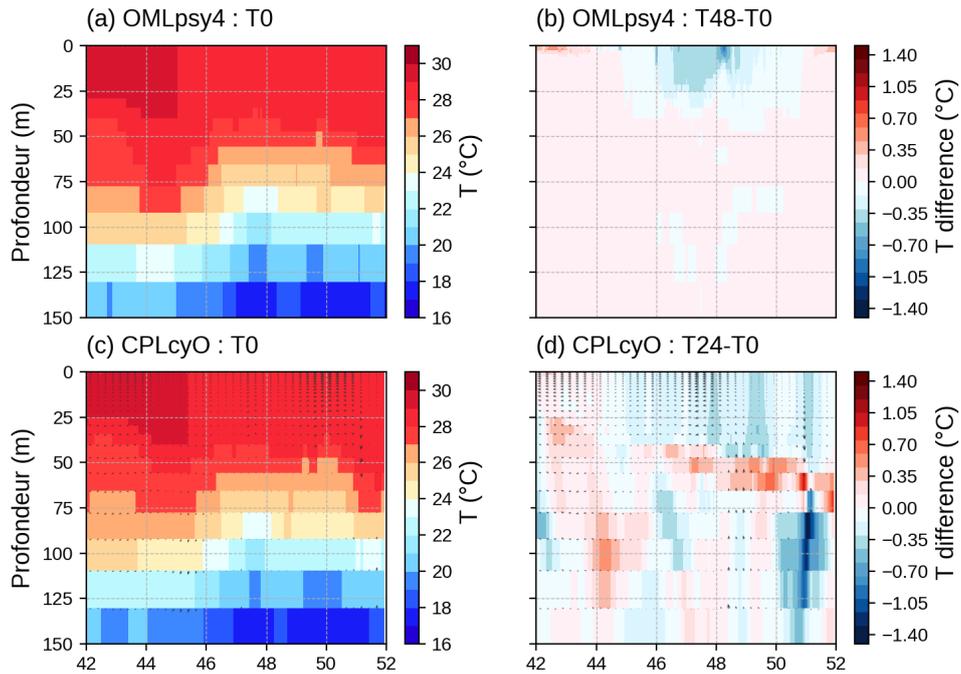


Figure 17. West-East cross-section along the dashed grey line on Figure 16 showing the evolution of the ocean temperature (T) in the run of the 6 December 2019 at 00 UTC (a-c) initial condition of simulation with the OMLpsy4 (top) and CPLcyO (bottom) configurations and (b-d) temperature (T) difference (°C) between the initial condition and the 48 h forecast

587 In the case of Belna, the impact of the 3D coupling with the ocean is limited. In
 588 particular, the lack of coupling was not responsible for the large intensity estimation in
 589 the last 36 h of the operational forecast starting on the 6 December 2019 at 00 UTC. The
 590 reason of this intensity error has yet to be understood with future works.

591 3.4 Batsirai: a case with buoy observations

592 We present here an analysis of TC Batsirai, which developed into a category 4 TC
 593 during the 2021-2022 TC season of the SWIO, and for which in situ buoy observations
 594 are available, unlike TCs of the previous three seasons. For this case, we mainly focus
 595 on the comparison of the oceanic state of the CPLcyO and OMLpsy4 experiment with
 596 the buoy measurements, taking advantage of such a rare opportunity in the SWIO basin
 597 to validate the evolution of the SST in the model in the vicinity of a TC with good qual-
 598 ity measurements.

599 **3.4.1 Batsirai analysis by the RMSC-La Réunion**

600 Batsirai was the second tropical system and the first to reach the TC level of the
601 2021-2022 cyclone season in the SWIO. It formed within a very active monsoon trough
602 on the 24 January 2022 in the northeast of the SWIO basin at the edge of a trans-equatorial
603 monsoon flow. After a short phase of rapid intensification on 27 January, Batsirai weak-
604 ens over the following days as it encounters larger wind shear and dry air intrusions along
605 its westward track. On 29 January, it meets much more favourable conditions and re-
606 intensifies into a Category 4 hurricane. As it travels north of Mauritius on 31 January,
607 it causes heavy flooding on the Island of St Brandon because of the combined effect of
608 a storm surge and a swell with the largest waves estimated between seven and nine me-
609 ters. Batsirai becomes an ITC on 2 February as it is the closest to La Réunion. Winds
610 exceeded 42 m s^{-1} in the highest ridges of La Réunion. Exceptional rainfall of about 1500
611 mm have been recorded on the volcano area during this episode. The storm underwent
612 an eyewall replacement the following day and fluctuated in intensity before making land-
613 fall on the East coast of Madagascar on 5 February as a category 3 cyclone causing at
614 least 121 casualties and a lot of damages on housing and infrastructures. The system then
615 weakened quickly as it crossed Madagascar. It emerged as a TS between Mozambique
616 and southern Madagascar on 7 February, and then became a post-tropical depression on
617 8 February.

618 **3.4.2 Comparison of TC Batsirai simulations with buoy measurements** 619 **and SAR winds**

620 As it is often the case in an operational context, the NWP guidances for TC tracks
621 and intensities are not perfect, but they still remain a valuable source of information for
622 the RSMCs. As Batsirai was at its closest from La Réunion, the uncertainty in the track
623 forecast was still large both in cross track and along track direction in most NWP mod-
624 els. AROME overestimated the speed of displacement and the Southward curvature of
625 Batsirai track. For the more detailed discussion below, we have selected the 48 h run from
626 the 3 February 2022 at 00 UTC as it covered the period when the TC displacement is
627 the slowest. With the criteria used for the TC of the 2018-2019 and 2019-2020 seasons
628 in the previous sections, Batsirai would be classified as a slow cyclone during this 48 h
629 period.

630 The initial intensity of the TC in AROME-IO is underestimated after the 6-h warmup
 631 (Figure 18a). In the operational configuration OMLpsy4 from the 3 February 2022 at
 632 00 UTC, the TC rapidly intensify and overtake the observed intensity after 12 hours of
 633 simulation (Figure 18b). In the coupled run CPLcyO, the intensification is weaker (Fig-
 634 ure 18) but the winds in the eyewall after 39 h of simulation are nevertheless 2-3 m s^{-1}
 635 stronger than the estimation given by the SAR wind retrieval (Figure 19). At the time
 636 of the SAR observation, the asymmetry of the system with weaker winds in the North-
 637 East quadrant is better seen in the coupled simulation than in the operational config-
 638 uration.

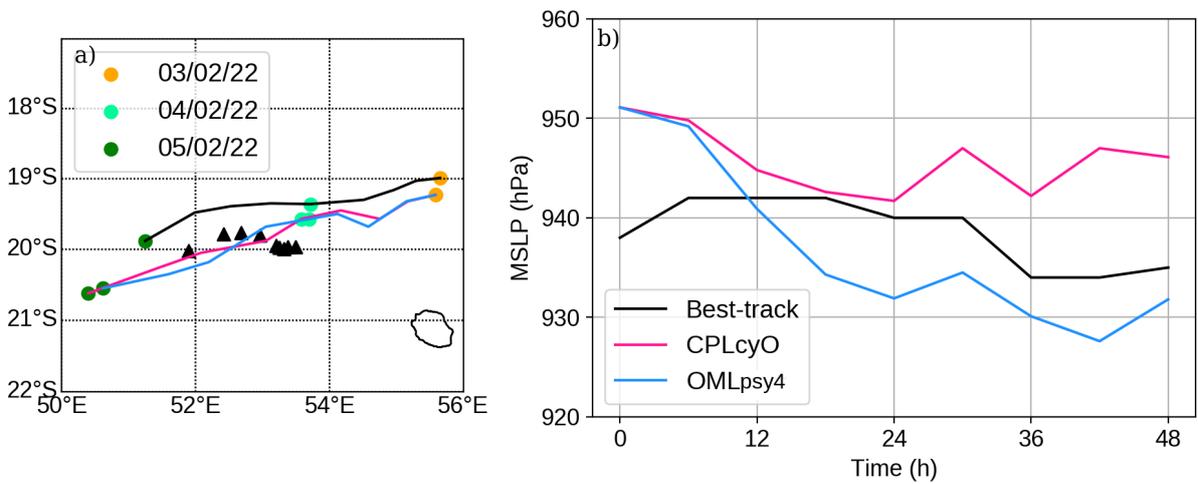


Figure 18. (a) Track of Batsirai and (b) Time evolution of Pmin (hPa) from the 3 February 2022 at 00 UTC to the 5 February 2022 at 00 UTC (48 h) for the RSMC La Réunion BT data (black) and the CPLpsy4 (pink) and OMLpsy4 (blue) configurations. The black triangles show the position of a drifting buoy every 6 h between the 3 and the 5 February 2022.

639 TC Batsirai triggers a SST cooling North of the Mascarene Islands when it displace-
 640 ment velocity is at its slowest. In the cold wake on the South side of Batsirai track sim-
 641 ulated by CPLcyO configuration, the maximum cooling is about 4°C (Figure 20). In the
 642 OMLpsy4, it is only about 1°C.

643 Figure 21 shows the evolution of the SST measured by the drifting buoy indicated
 644 by a black triangle on Figure 20 (black line). The SST at the buoy loses about 3°C be-
 645 tween the 3 February 2022 at 00 UTC and the 5 February 2022 at 00 UTC. The pink
 646 curves on Figure 21 show the evolution of the SST at the position of the buoy in three

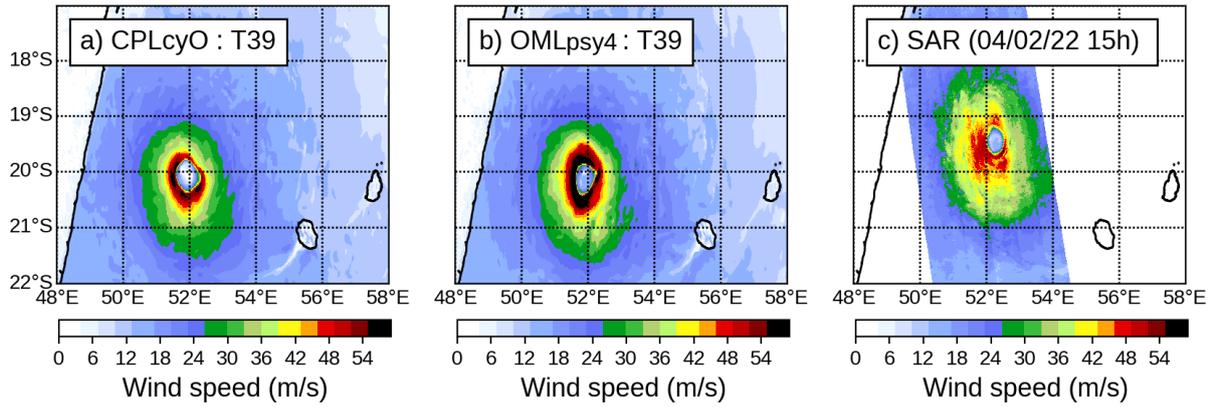


Figure 19. TC Batsirai surface winds: (a-b) 10 m wind speed (m s^{-1}) of the 39 h forecasts verifying on 4 February 2022, 15 UTC for both CPLcyO and OMLpsy4 c) SAR surface wind speed (m s^{-1}) on 4 February 2022 at 15h.

647 successive CPLcyO runs (2 February at 00UTC, 3 February at 00UTC and 4 February
 648 2022 at 00 UTC). AROME-NEMO reproduces a very similar evolution of the SST but
 649 with an advance of about 24 h. The error in the track position, and especially the along
 650 track error that results from an overestimation of the TC translation velocity is a likely
 651 reason for the timing error of the SST cooling in the model in this case.

652 This last case illustrates well the difficulty of the comparison of TC simulations with
 653 observations, especially when there is an error in the TC position forecast. However, the
 654 confrontation of the buoy observations with the result of the CPLcyO simulation shows
 655 that the oceanic response of the coupled system AROME-IO/NEMO is realistic when
 656 the surface winds of the TC are close to the observed values.

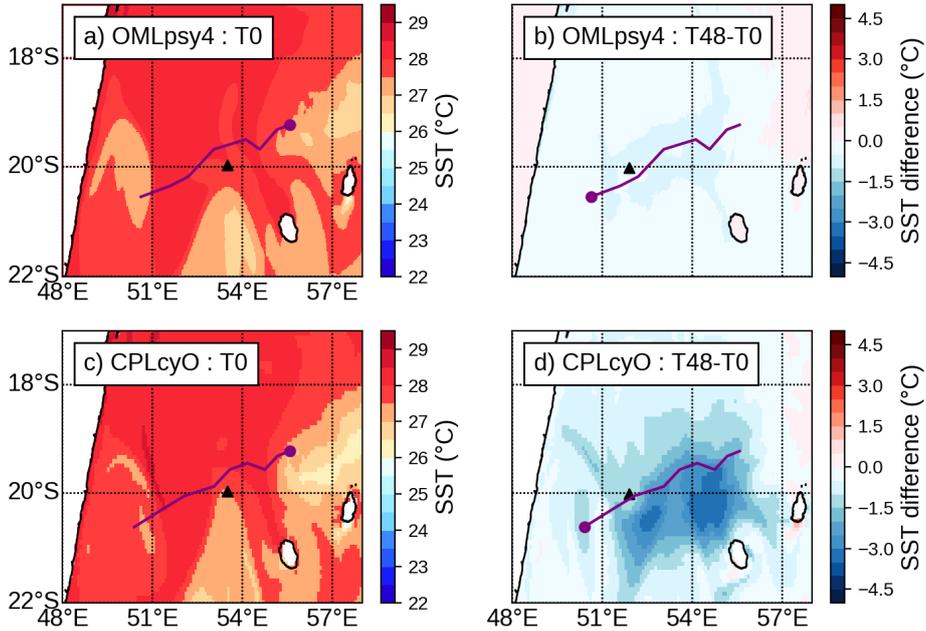


Figure 20. SST ($^{\circ}\text{C}$) in the vicinity of Batsirai for the simulation starting at 00 UTC on the 3 February 2022 : (a-c) initial condition of simulation with the OMLpsy4 (top) and CPLcyO (bottom) configurations and (b-d) SST difference ($^{\circ}\text{C}$) between the initial condition and the 48 h forecast. The purple point gives the position of the cyclone at the time of the figure and the track of the cyclone is represented by the purple line. The black triangle show the position of the drifting buoy.

657 **4 Discussion and conclusion**

658 The overseas version of the convection-permitting numerical weather prediction model
 659 AROME-IO which is used by the RSMC-cyclones of La Réunion, is coupled every 5 min-
 660 utes to a 1D parameterization of the OML. The present study aimed at testing the re-
 661 placement of the simplified OA interaction described by the 1D parameterization of the
 662 OML by a full 3D coupling with the oceanic model NEMO in a future configuration of
 663 AROME-IO. The implementation of the coupled configuration AROME-IO/NEMO which
 664 was the first step needed for this study has then been facilitated by the developments
 665 which had already been made in the surface platform SURFEX used by AROME. The
 666 data exchange between the atmosphere and the ocean is managed by the OASIS cou-
 667 pler every hour. Unlike the 1D parameterization, the ocean model needs boundary con-

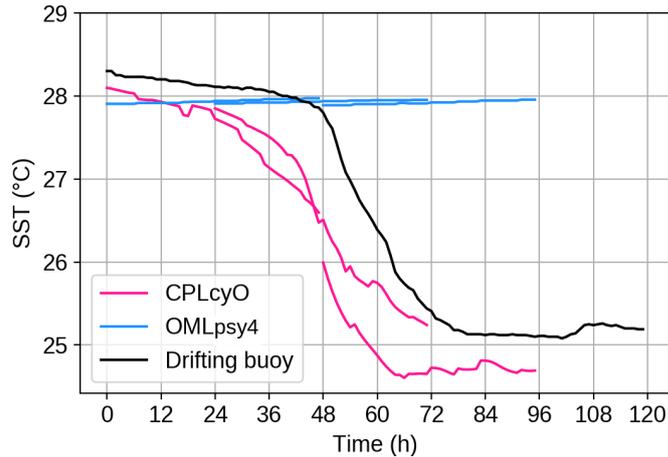


Figure 21. Time evolution of the SST ($^{\circ}\text{C}$) measured by drifters in the vicinity of Batsirai track (black line). Comparison with SST at the location of the drifters in 3 successive 48 h simulations starting on the 2 February, 3 February and 4 February 2022 at 00UTC, with the CPLcyO configuration (pink) and the OMLpsy4 configuration (blue).

668 conditions which are prepared from the global oceanic products of MERCATOR-Ocean. In
 669 the design of the coupled system, we kept in mind the constraints imposed by a future
 670 operational system. For example, we only used data sources which would be available
 671 in real time. Early on in the design process, the question of the initialisation of the ocean
 672 has been raised for the case of a forecast system running 4 times a day for short fore-
 673 casts (48 to 78 hours). Several solutions have been tested and two of them have been se-
 674 lected for the analyses presented in this article.

675 Atmospheric and oceanic in situ observations in the vicinity of TCs are very sparse
 676 in the IO. High resolution satellite wind measurements are irregular and SST products
 677 are of low quality due to cloud cover. Unlike in the Atlantic or North Pacific basin, it
 678 is difficult to validate NWP configuration against observation. The validation of TC fore-
 679 cast is then often limited to the comparison of macroscopic characteristics such as po-
 680 sition, minimum of pressure at the centre of the TC and maximal wind. In a few cases,
 681 high resolution SAR images and buoys measurement give useful information for model
 682 evaluation.

683 The main results of this study are based on a comparison between five configura-
 684 tions of AROME-IO; one with constant SST, two with the coupled AROME-NEMO model

685 and two with the OML parameterization. For each of these two OA coupling options,
 686 the initial condition of the ocean is either directly interpolated from the MERCATOR-
 687 Ocean PSY4 products as it is the case for the initialisation of the OML in the current
 688 operational model or it is cycled by the AROME-NEMO system since a few days before
 689 the cyclone intensification. Three-day forecasts for about 30 starting dates and 7 differ-
 690 ent TCs of the 2018-2019 and 2019-2020 TC cyclone seasons have been produced. The
 691 simulations are first compared with classical scores against the RSMC BT data. Then,
 692 case studies illustrate the differences between the different coupling solution for a selec-
 693 tion of 3 TCs.

694 The scores shows that:

- 695 • there is very little impact of the 3D coupling on the TC track. This was expected
 696 as, in a limited area model, the track is mainly driven by the larger scale of the
 697 host model.
- 698 • there is an improvement of the cyclone intensity forecast with the 3D coupling in
 699 AROME-IO both in terms of bias and standard deviation. These improvements
 700 are particularly true for TCs that encounter a slow propagation phase (less than
 701 2 m s^{-1} for at least 12 hours).
- 702 • the memory of the mesoscale OA interaction also contributes to better TC inten-
 703 sity in the first 36 h of the forecast, both for 3D coupling and 1D OML param-
 704 eterization.

705 These results are confirmed by the 3 case studies. The detailed analyses of the OA
 706 interaction also show that:

- 707 • very intense winds in a stationary TC may trigger strong upwelling which are cool-
 708 ing the ocean surface of more than 5°C per day. In such extreme cases, the OA
 709 thermodynamic fluxes are reversed compared to the usual TC configuration where
 710 the ocean fuels the TCs. The TC intensification is then significantly affected. In-
 711 tense coolings modify the structure of the boundary layer, both in the wall and
 712 in the eye, with possible formation of low level clouds in the eye.
- 713 • the 3D upwellings are deep circulations which affect first the bottom of the OML.
 714 The strongest upwelling will completely modify the structure of the OML and the
 715 well mixed water will be replaced by cold water advected upward across the ther-

716 mocline. In case of regularly moving TC, weak upwellings are triggered; colder wa-
717 ter accumulates under the thermocline and slowly mixes with the upper layer. The
718 feedback on the TC intensity is then weak as the maximum cooling is weaker and
719 anyway, the TC is not affected by the maximum cooling as it has moved on.

720 Even if the 3D coupling reduces the tendency of AROME-IO to over intensify al-
721 ready intense cyclones, the scores against the BT and the case studies show that the lim-
722 itation of the simplified 1D coupling with the ocean is only one aspect of the problem.

723 One other avenue to control the TC intensification is to improve the parameter-
724 ization of the surface fluxes in case of strong winds. In the ECUME "bulk" parameter-
725 ization of the surface fluxes above the ocean currently used in AROME, the momentum
726 flux decreases for winds larger than 30 m s^{-1} and then levels off for larger wind speeds.
727 However, there is no real consensus about the behaviour of the momentum and heat fluxes
728 for very strong winds (and almost no measurements). A possibility would be to move
729 from a strongly parameterize "bulk" scheme as ECUME towards a scheme with more
730 degrees of freedom. We are currently testing the WASP scheme (Wave-Age Stress de-
731 pendent Parameterization; Sauvage et al., 2020) which has recently been implemented
732 in SURFEX. WASP is based on the COARE3.0 (Fairall et al., 2003) and COARE 3.5
733 (Edson et al., 2013) schemes. It offers the possibility to explicitly account for the wave
734 growth in the calculation of the roughness length at the wind range where the momen-
735 tum of the atmosphere transferred to the waves is between 7 and 25 m s^{-1} . Above 25
736 m s^{-1} , the contribution of wave breaking is dominant and the wave age is no longer a
737 sufficient parameter to represent the impact of the sea state on the surface roughness.
738 The contribution of sea sprays to the OA exchanges in case of high winds is also prob-
739 ably to be considered. Several ongoing researches on this topics may contribute to fur-
740 ther improvement of the air-sea exchange parameterization in TC conditions.

741 A second avenue would be to work on the microphysics scheme. The ICE3 micro-
742 physics currently used in AROME is a 1 moment microphysics without any direct feed-
743 back on aerosol concentration. Tests with a 2 moment microphysics have shown that a
744 prognostic concentration in marine and dust aerosol limits the TC growth (T. Hoarau
745 et al., 2018). However, the introduction of new degrees of freedom in the system brings
746 new sources of uncertainty and should then be very carefully evaluated in a prospective
747 of operational use. The initialisation and the lateral boundary coupling of the aerosol

748 and the parameterization of the sources of aerosol are also factors that must be of suf-
749 ficient quality for the performance of the models to improve.

750 AROME-NEMO is an evolution toward a more realistic regional forecast system
751 for TCs. With a resolution of 1/12 of a degree, the computational cost of the ocean model
752 NEMO remains negligible compared to the cost of the atmospheric model AROME. There
753 is however a clear increase in the complexity of the forecast suite as the suite needs to
754 prepare the initial and lateral boundary conditions for NEMO and the data to feed the
755 OASIS coupler at the beginning of the simulation.

756 This study also illustrates the importance of the ocean initial conditions in a cou-
757 pled NWP suite. We have found that the MERCATOR PSY4 products which are cur-
758 rently forced by winds with a native resolution of about 10 km show a much weaker re-
759 sponse in term of upwelling (but also in term of quasi-inertial waves; G. Hoarau & Malardel,
760 2021) than what we have found in the coupled AROME-NEMO configuration. Our re-
761 sults show that a suite where the ocean model is initialised every 24 hours by a new PSY4
762 ocean state loses the memory of the TC high resolution wind forcing. We have also shown
763 that the memory of the high wind forcing can be kept if the ocean state from a previ-
764 ous forecast is used to initialise the next forecast. However, in an operational suite, the
765 ocean cannot be cycled indefinitely without any correction towards oceanic observation.
766 It is then important to implement a solution which combine the PSY4 ocean state which
767 is regularly updated by an oceanic data assimilation and a previous forecast of the ocean
768 which has seen high resolution winds. Several solutions from a simple linear relaxation
769 toward the PSY4 state to more scale selective nudging procedures will be tested.

770 In this study, we made the choice to use the same atmospheric initial condition for
771 all the experiments in order to focus on the sensitivity to the oceanic component and the
772 coupling solution. But an effort to improve the initial balance between the atmosphere
773 and the ocean in the initial conditions of both fluids will be needed. The simplest so-
774 lution will be to adapt the warmup procedure which is currently used for the atmosphere
775 to reduce the initial spin-up of the downscaling adjustment of the IFS fields to the AROME
776 resolution to the coupled configuration. In parallel, research and development activities
777 towards a coupled data assimilation system for AROME are starting in the research groups
778 at Météo-France.

779 **Acknowledgments**

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781 **Availability Statement**

782 The atmospheric model used in this study is AROME Cycle 43t2 which has been
 783 the operational version of the AROME-overseas from July 2019 to June 2022. The ocean
 784 model is NEMO, version 3.6. The atmospheric and the oceanic models are coupled with
 785 the OASIS3-MCT coupler.

786 The SAR data have been uploaded from <https://cyclobs.ifremer.fr/app/> and the
 787 data from the ARGO drifters from <http://www.coriolis.eu.org/Data-Products/Data-selection>.

788 The Best Track data have been extracted from the Best Track data base of the Di-
 789 rection Régional de l’Océan Indien (DIROI) of Météo-France. These data are shared with
 790 the IBTracs data base (<https://www.ncdc.noaa.gov/ibtracs/>) after a subjective reanal-
 791 ysis by the DIROI forecasters at the end of each TC season.

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