

Dynamical contribution to sea surface salinity variations in the eastern Gulf of Guinea based on numerical modelling

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Abstract:

In this study, we analyse the seasonal variability of the sea surface salinity (SSS) for two coastal regions of the Gulf of Guinea from 1995 to 2006 using a high resolution model (1/12°) embedded in a Tropical Atlantic (1/4°) model. Compared with observations and climatologies, our model demonstrates a good capability to reproduce the seasonal and spatial variations of the SSS and mixed layer depth. Sensitivity experiments are carried out to assess the respective impacts of precipitations and river discharge on the spatial structure and seasonal variations of the SSS in the eastern part of the Gulf of Guinea. In the Bight of Biafra, both precipitations and river runoffs are necessary to observe permanent low SSS values but the river discharge has the strongest impact on the seasonal variations of the SSS. South of the equator, the Congo river discharge alone is sufficient to explain most of the SSS structure and its seasonal variability. However, mixed layer budgets for salinity reveal the necessity to take into account the horizontal and vertical dynamics to explain the seasonal evolution of the salinity in the mixed layer. Indeed evaporation, precipitations and runoffs represent a relatively small contribution to the budgets locally at intraseasonal to seasonal time scales. Horizontal advection always contribute to spread the low salinity coastal waters offshore and thus decrease the salinity in the eastern Gulf of Guinea. For the Bight of Biafra and the Congo plume region, the strong seasonal increase of the SSS observed from May/June to August/September, when the trade winds intensify, results from a decreasing offshore spread of freshwater associated with an intensification of the salt input from the subsurface. In the Congo plume region, the subsurface salt comes mainly from advection due to a strong upwelling but for the Bight of Biafra, entrainment and vertical mixing also play a role. The seasonal evolution of horizontal advection in the Bight of Biafra is mainly driven by eddy correlations between salinity and velocities, but it is not the case in the Congo plume.

Keywords: Gulf of Guinea ; Salinity ; Modelling ; Congo ; Bight of Biafra ; Mixed layer budget

** This paper is a contribution to the special issue on tropical Atlantic variability and coupled model climate biases that have been the focus of the recently completed Tropical Atlantic Climate Experiment (TACE), an international CLIVAR program (<http://www.clivar.org/organization/atlantic/tace>). This special issue is coordinated by William Johns, Peter Brandt, and Ping Chang, representatives of the TACE Observations and TACE Modeling and Synthesis working groups.*

1 Introduction

The Gulf of Guinea is a region of importance for the development of the African Monsoon (Redelsperger et al. [2006](#)) because of the ocean/atmosphere interactions.

39 The existence of a shallow thermocline and mixed layer (de Boyer Montegut et al
40 2004) in the eastern part of the Tropical Atlantic, that can be easily eroded, is
41 one of the reasons for the existence of the Atlantic cold tongue which is a key
42 feature for the formation of the monsoon. Stratification is thus one of the elements
43 which must be understood to better describe the fluxes and interactions between
44 the tropical ocean and the atmospheric boundary layer. Already sharp because of
45 the thin thermocline, the stratification in the central and eastern part of the Gulf
46 of Guinea is reinforced by a strong halocline due to the presence of anomalously
47 freshwaters extending from the eastern coast to 0° E or even farther west (Dessier
48 and Donguy 1994), and a subsurface salinity maximum due to subtropical waters
49 advected by the Equatorial undercurrent (Blanke et al 2002).

50 In the Bight of Biafra north of the Equator and offshore Angola and Gabon
51 south of it, the low salinity values observed in the surface layer (lower than 31 psu
52 in the north and 32 psu in the south, Figure 1) result from high precipitations and
53 river discharge. The importance of the salinity gap between these water masses
54 and the tropical surface water (close to 35.5 psu, Stramma and Schott (1999)), can
55 be easily explained by the amount of freshwater concerned. The Congo river is the
56 second most important in the world with an average discharge of 40 mSv (Mahé
57 and Olivry 1999) and the Niger river is the twelfth with 7 mSv (Dai and Trenberth
58 2002), both with large seasonal variations. Precipitations over the whole Gulf of
59 Guinea are substantial (140 mSv using the dataset of Large and Yeager (2009))
60 and also vary seasonally.

61 A large number of studies has been devoted to the variability of temperature in
62 the mixed layer of the tropical Atlantic (see Giordani et al (2013), Hummels et al
63 (2012) or Jouanno et al (2011) for recent examples). However, in situ observations

64 of salinity have a sparse spatial and temporal resolution compared with temper-
65 ature (Reverdin et al 2007). Remote sensing of salinity has become possible very
66 recently, with large uncertainties still (Tzortzi et al 2013). Da-Allada et al (2013a)
67 recently computed a budget of mixed layer salinity from in situ observations in
68 the whole tropical Atlantic. However, the sparseness of the data makes the results
69 questionable for the coastal regions of the eastern Gulf of Guinea where the lowest
70 salinity waters are found.

71 In this paper we attempt to better understand the mechanisms which drive
72 the seasonal variations of the SSS of the eastern Gulf of Guinea, concentrating
73 on the Bight of Biafra and the Congo plume regions. Our objectives are : 1) to
74 determine the exact contribution of the precipitations and river outflow in term
75 of mean state and seasonal variations of the SSS, 2) to determine the dynamical
76 contributions to the seasonal cycle and especially the importance of horizontal and
77 vertical processes.

78 As in-situ observations for the salinity in these regions are not sufficient to make
79 a complete analysis of the seasonal variations of the SSS, we choose to use numer-
80 ical modelling to assess the mechanisms corresponding to these two questions.
81 Regarding the impact of freshwater sources, our analyses are based on sensitivity
82 experiments on the freshwater forcing. We evaluate the key mechanisms based on
83 the diagnostics introduced by Vialard et al (2001) for mixed layer temperature,
84 but applied here to the seasonal mixed layer budget for salinity.

85 This paper is organized as follow. Section 2 describes the characteristics of our
86 regional model. Section 3 presents a discussion about the impact of precipitations
87 and river runoffs in term of spatial repartition of the SSS and mean amplitude of
88 the seasonal variations. In section 4 we quantify the importance of various physical

89 processes using the mixed layer budget for salinity and relate them to the regional
90 dynamics. Finally, discussions and concluding remarks are presented in section 5.

91 **2 Numerical model and validation**

92 2.1 Model characteristics

93 Because of sparse spatial and temporal resolution of in-situ data for SSS, a regional
94 ocean model is set up in order to represent the oceanic processes in the Gulf of
95 Guinea. We use the NEMO 3.2.1 numerical model (Madec 2008) with AGRIF
96 online refinement to combine $1/4^\circ$ and $1/12^\circ$ grids with two ways interactions
97 (Debreu and Blayo 2008). Our configuration is based on the $1/4^\circ$ global experi-
98 ment ORCA025.L75 developed by the DRAKKAR team (Barnier et al 2006) and
99 the regional $1/12^\circ$ configuration used by Guiavarc'h et al (2008). The domain cov-
100 ers the Tropical Atlantic (from 30°S to 30°N and from 60°W to 15°E) with a $1/4^\circ$
101 grid and the Gulf of Guinea with a $1/12^\circ$ one (AGRIF zoom from 10°W to 15°E
102 and from 15°S to 8°N), these grids can be seen on Figure 1. The $1/12^\circ$ resolution
103 is chosen for the eastern Gulf of Guinea because currents on the continental slope
104 are too weak compared with observations at $1/4^\circ$ (Guiavarc'h et al (2008), their
105 Figure 7). Both grids have 75 vertical levels in partial steps with a first layer of
106 1 meter thickness. We use extractions of the global $1/4^\circ$ and $1/12^\circ$ bathymetries
107 built by Mercator Ocean (<http://www.mercator-ocean.fr>). Radiative open bound-
108 aries (Treguier et al 2001) are set up in the eastern, southern and northern limits
109 of the $1/4^\circ$ grid. They radiate perturbation outward and relax the model variables
110 to 5 day averages of the ORCA025.L75 global experiment.

111 An energy-ensrophy conserving momentum advection scheme (Penduff et al
112 2007) is used for the dynamics. Lateral diffusion of momentum is done with a
113 horizontal bilaplacian operator with coefficient $1.5e^{11} \text{ m}^4.\text{s}^{-2}$ in the $1/4^\circ$ grid and
114 $1.2e^{10} \text{ m}^4.\text{s}^{-2}$ in the $1/12^\circ$ grid. The time steps are 2400s and 800s for the $1/4^\circ$
115 and $1/12^\circ$ grids respectively. The advection of passive tracers is based on a Total
116 Variance Dissipation (TVD) scheme and diffusion is parameterized by a laplacian
117 isopycnal operator with coefficient $300 \text{ m}^2.\text{s}^{-1}$ ($1/4^\circ$) and $100 \text{ m}^2.\text{s}^{-1}$ ($1/12^\circ$). The
118 vertical diffusion coefficient is similar for the two grids and is given by a Turbulent
119 Kinetic Energy (TKE) second order closure scheme (Blanke and Delecluse 1993).
120 An enhanced vertical diffusion of $1 \text{ m}^2.\text{s}^{-1}$ is applied on tracer and momentum in
121 case of static instability. The Agrif coupling between our grids uses a laplacian
122 diffusivity in the sponge layer equal to $300 \text{ m}^2.\text{s}^{-1}$ for tracers and dynamics. The
123 baroclinic update between grids is done at each time step of the mother grid. The
124 model starts from a climatology of temperature and salinity (Levitus 1986) in 1990
125 and is integrated to 2006. To perform all the diagnostics done in this paper, we
126 use 5 day averages for the period from 1995 to 2006.

127 The atmospheric forcing at the surface is computed with the CORE bulk formu-
128 lation (Large and Yeager 2004). We use a composite forcing based on the DFS4.3
129 forcing (Brodeau et al 2010) set up by the DRAKKAR team and the ERA-interim
130 forcing from the ECMWF. From DFS4.3 we use observed precipitations and solar
131 radiation (based on satellite observations from the dataset of Large and Yeager
132 (2009); precipitations are based on the Global Precipitation Climatology Project,
133 GPCP). From ERA-interim we use temperature, humidity and winds at 2 meters.
134 The model takes into account the diurnal cycle on solar radiation. The short wave

135 radiation penetration depends on the ocean colour based on a SeaWifs climatology,
136 so the extinction coefficients vary horizontally (Madec 2008).

137 River runoffs are prescribed by a surface freshwater flux near the river mouth
138 and along the coast. Coastal runoff values come from the inter-annual dataset of
139 Dai et al (2009) based on in-situ measurements and model reconstructions with
140 a River Transport Model (RTM) over the period 1990-2004. The focus of this
141 study being the seasonal cycle, our reference experiment (hereafter REF) is run
142 with climatological runoffs. Another experiment (D09) is run with inter-annually
143 varying runoffs. The climatological runoff values for REF are the averaged fields
144 of D09 from 1990 to 2004. Indeed, for the period considered here, the use of the
145 Dai and Trenberth (2002) climatology implies an important overestimation of the
146 river runoffs relative to the inter-annual dataset, because of a decreasing inflow in
147 this region since 1948 (equal to 15 % from Dai et al (2009) and Mahé and Olivry
148 (1999)). The comparison of the spatially averaged SSS between the REF simulation
149 and D09 (Figure 2a) shows that the inter-annual variations of the river runoffs do
150 not impact much the SSS variability. Especially, the inter-annual SSS anomalies
151 of these two experiments are practically equal (Figure 2b). Finally, as Ferry and
152 Reverdin (2004) demonstrate that the only simulation they perform which is able
153 to reproduce the inter-annual variability in the western Tropical Atlantic is the
154 one with no SSS restoring, we do not use surface restoring for salinity to avoid
155 excessive damping of the inter-annual variability.

156 2.2 Validation of the reference experiment

157 2.2.1 *Surface salinity*

158 Our model is comparable to those used by Peter et al (2006) and Jouanno et al
159 (2011), and shares many of their characteristics concerning the heat content in the
160 surface and subsurface layer. Here we focus the validation on the salinity field for
161 our REF experiment; more details are found in Berger (2012). The comparison
162 between the SSS of our REF experiment and the ARV09 climatology from Gail-
163 lard et al (2009) demonstrates the good capability of the model to reproduce the
164 three main features of the SSS in the Gulf of Guinea (Figure 3). First, the south-
165 westward salinity gradient at regional scale appears similar in the climatology and
166 the model, the latter being less smooth because of its better spatial resolution.
167 Second, the large plumes and desalinated waters in the Bight of Biafra (with the
168 Niger and some important rivers) and offshore Gabon and Angola between 8 and
169 4°S (with the Congo River) appear positioned correctly. Finally, desalinated wa-
170 ters north and south of the equator are separated by higher salinity water offshore
171 the Cap Lopez, near 1°S, with salinity equal to 36.5 psu for the climatology and
172 35 psu for the REF model.

173 However, the salinity along the coast is lower in the model than in the clima-
174 tology, with differences up to 3 psu in the Congo plume and more than 7 psu in the
175 Bight of Biafra. This discrepancy is at least partly due to the lack of observations
176 near the coast (see Figure 1 of Da-Allada et al (2013a)) and the low resolution
177 of the climatology: 0.5°, with a smoothing radius of 300km (Gaillard et al 2009).
178 Note that new satellite observations (Tzortzi et al 2013) show a lower salinity

179 along the coast than the climatology, and a stronger connection between the low
180 salinity waters of the Congo and the Bight of Biafra, in agreement with the model.

181 The SSS from our REF experiment is also compared in Figure 4 to the available
182 SSS data from the PIRATA mooring at 0°S - 0°E and also with the short record
183 (less than one year) at 6°S - 8°E (Bourlès et al 2008). Like the SST, the SSS in the
184 Gulf of Guinea presents a strong seasonal cycle along the equator and offshore the
185 coast at 6°S (Dessier and Donguy 1994; Eisma and Van Bennekom 1978). Along
186 the equator, the model reproduces the seasonal cycle quite well, albeit with a larger
187 seasonal amplitude (about 2 psu in the model and closer to 1.8 psu in the data).
188 The model shows a phase shift with respect to the observations for the salinization
189 phase in spring for years 2001 and 2006, but not for the other years. At 6°S the
190 model seems to underestimate the decrease in salinity observed in november 2006.

191 *2.2.2 Stratification and mixed layer depth*

192 The stratification of our model in the Gulf of Guinea is too strong compared
193 with the ARV09 climatology, for both temperature and salinity (Figure 5). If the
194 thermocline appears a little bit too sharp, the main concern about stratification
195 results from the salinity between 0 and 60 meters depth. Indeed, the model is
196 0.5 psu fresher than the climatology at the surface (this is due to lower values
197 near the coast as shown in Figure 3) and 0.15 psu saltier at 40 m depth. In the
198 model, the salty waters carried by the Equatorial undercurrent below the surface
199 layer in the Gulf of Guinea are not sufficiently eroded by mixing with the overlying
200 freshwaters in comparison with observations (Kolodziejczyk et al 2013). In setting
201 up the model, we have tried to adjust vertical mixing parameters (such as the
202 background viscosity and diffusivity of the TKE mixing model) but we have not

203 been able to improve this model bias. This strong vertical gradient in the model
204 can influence the exchanges between the surface and subsurface layers.

205 For the mixed layer budgets, the capability of the model to reproduce correctly
206 the spatial structure of the mixed layer and its temporal evolution is of particular
207 importance. Following the recommendations of de Boyer Montegut et al (2004) for
208 tropical regions, we use a 0.03 kg.m^{-3} density criterion to define the mixed layer
209 depth in the model. The mean state of the mixed layer in our REF experiment
210 is compared to the de Boyer Montegut et al (2004) climatology, using the 2008
211 update, in Figure 6. Both climatology and REF present shallow mixed layer depths
212 along the eastern coast, particularly in the Bight of Biafra and in the region of
213 the Congo plume with an averaged depth of 10 to 12 meters. Both also depict
214 a deepening of the mixed layer in the region where the Guinea Current flows,
215 around 2°N , between the western boundary and 5°E . In this region, the mixed
216 layer depth is equal to 26 m on average. Finally, both model and observations
217 present a shallower mixed layer along the equator than in the surrounding regions;
218 however, this shallowing is exaggerated in the model. This can be explained by
219 the strong stratification (Figure 5) which occurs along the equator more than in
220 the rest of the basin. A similar discrepancy happens along the southern coast,
221 between 13 and 6°S , with a very shallow mixed layer in the model compared to
222 the climatology. However, the lack of observations in this region make it difficult
223 to conclude that the model is deficient in this area. The temporal variations of the
224 mixed layer will be discussed in section 4.2 and are thus not presented here.

225 **3 freshwater forcing and SSS variability**

226 3.1 freshwater input in the eastern Gulf of Guinea

227 Table 1 provides the respective volumes of precipitations and river runoffs in our
228 two regions of interest, the Bight of Biafra and the Congo plume (black boxes on
229 Figure 1) as well as for the Gulf of Guinea domain covered by our $1/12^\circ$ grid.
230 The two regions concentrate 87% of the 80 mSv discharged by rivers in the Gulf
231 of Guinea, but they receive only 30% of the 140 mSv of precipitations over the
232 region.

233 In the Bight of Biafra, both precipitations and river runoffs present the same
234 mean volume flux, equal to 27 mSv. South of the equator, the Congo represents
235 76% of the freshwater discharge (44 mSv against 14 mSv for precipitations) and
236 even the minimum discharge that occurs in August (33 mSv) is higher than the
237 maximum discharge due to precipitations (31 mSv during April). In addition, the
238 seasonal variations of each source of freshwater differ, depending on the region
239 (Figure 7): semi annual for precipitation and river runoffs in the Bight of Biafra,
240 semi annual for the Congo discharge and annual for precipitations south of the
241 equator.

242 In the Biafra box in the northern hemisphere, the semi annual evolution of
243 both precipitations and river runoffs (Figure 7) is associated with the African
244 monsoon. Indeed during April, the maximum of precipitations is related to the
245 northern displacement of the Inter Tropical Convergence Zone (ITCZ) which moves
246 from the ocean to sub Saharan regions over the continent (Philander et al 1996).
247 When the ITCZ goes back to its most southerly position over the ocean around
248 November, a second intensification of precipitations occurs (Redelsperger et al

249 2006). In between, around August, precipitations over the ocean are minimum
250 when the monsoon front is in its most northerly position. Due to the time needed
251 for precipitations over the continent to reach the ocean, the maximum runoff occurs
252 five months later, during September/November. Contrary to the precipitations
253 over the ocean, precipitations that cover the river catchment area present only
254 annual variations (Mahé and Olivry 1999), which explain the weaker semi annual
255 cycle of the runoffs compared with precipitations.

256 In the Congo box, the seasonal variation of the runoffs is relatively weak.
257 Indeed, due to its huge catchment area, which covers both hemisphere in the central
258 Africa (Laraque et al 2001), the Congo river is influenced by rainfall seasons in
259 both hemispheres. As a consequence, it has always a part of its basin under high
260 precipitations, which explains its important discharge equal to 44 mSv on average.
261 Because of the alternation in the rainfall seasons, the seasonal cycle of the river
262 appears semi-annual, with a maximum discharge of 60 mSv occurring in December
263 and a weaker relative maximum in May. Precipitations over this region follow an
264 annual cycle and are very weak from May to October.

265 3.2 Sensitivity experiments on freshwater forcing

266 To clarify the respective influence of precipitations and river runoffs on the spatial
267 structure and variations of the surface salinity we have performed two sensitivity
268 experiments. Based on our REF experiment, we perform a first simulation forced
269 by precipitation only (PRECIP) where river runoffs have been turned off. In the
270 second simulations (RUNOFF), precipitations have been turned off and it is thus
271 forced by river discharges only. As we remove a large part of the freshwater input

272 in these sensitivity tests, they are subject to a larger drift relative to the observed
273 climatology. An adjustment is necessary during the first years of these runs to
274 eliminate the low salinity water masses in the Bight of Biafra and offshore Angola
275 that cannot be maintained with only a part of the observed freshwater input. To
276 speed up this adjustment, we use a new initial state where the salinity in the
277 surface layer in the Bight of Biafra and offshore Angola is set to the mean value
278 of the SSS in the Gulf of Guinea. In addition, each simulation is integrated twice
279 longer than REF by repeating the forcing from 1990 to 2006. Only the second
280 integration is analysed.

281 Figure 8 shows the mean SSS from 1995 to 2006 for the PRECIP and RUNOFF
282 experiments, with only precipitations (a) and river runoffs (b) respectively. The
283 mean SSS on this figure must be compared with the mean SSS of our REF simu-
284 lation on Figure 3b to better appreciate the influence of each source of freshwater.
285 First, the meridional structure of the SSS, visible on both the climatological data
286 and our REF simulation (Figure 3) can be explained by the meridional structure
287 of the precipitations as already noticed by Yoo and Carton (1990). However, the
288 PRECIP simulation clearly demonstrates that the desalinization in the Bight of
289 Biafra is partly due to the large amount of precipitations discharged in this region
290 (27 mSv on average). Our test shows that the contribution of the precipitations
291 to the salinity anomaly can reach 3.5 psu, to compare with the 7.5 psu of the
292 REF case (differences between the Bight and the open ocean around 5°E). This
293 test also confirms the negligible role of precipitations south of the Equator as no
294 desalinization can be observed in the Congo box, as expected from Table 1. Sec-
295 ond, the RUNOFF simulation demonstrates the importance of the river runoffs
296 to explain both the spatial structure of the SSS in the eastern part of the Gulf

297 of Guinea and the amplitude of the desalinization compared to the open ocean.
298 River runoffs contribute for 4 psu in the Bight of Biafra. As expected, they explain
299 more than 90% of the desalinization in the Congo box, with an amplitude equal
300 to 7.5 psu (close to REF). In addition, even without precipitation to support the
301 desalinization south of the equator, we find that the Congo plume can spread to
302 5°E, practically the same extension as in the REF case.

303 Let us study the seasonal variability of SSS for each of the sensitivity experi-
304 ments. A mean seasonal cycle is constructed by computing 12 monthly means over
305 the period 1995-2006, and the seasonal amplitude is estimated as the difference
306 between the maximum and minimum monthly salinity at each location. A map of
307 the seasonal amplitude for our REF experiment is shown in Figure 9a. The highest
308 variability occurs along the coast (up to 8 psu), where huge river discharges take
309 place : in the Bight of Biafra north of equator and in the region of the Congo
310 plume. Indeed, this map of seasonal amplitude has guided our choice of the target
311 regions presented in Figure 1. The maximum seasonal amplitude is larger in the
312 model compared with the in-situ climatology, but it is close to the amplitude re-
313 vealed by the first satellite observations of SSS (Tzortzi et al 2013). The seasonal
314 variability decreases rapidly with the distance to the coast. Indeed, offshore of a 2
315 to 3° width band along the coast, the seasonal amplitude reaches only 2 psu. The
316 maximum of variability takes place in the Congo plume region, where the desalin-
317 ization is based upon only one source of freshwater and where the highest input is
318 concentrated on a small region. Despite a lower river discharge inflow (Table 1),
319 the region of high variability around the Bight of Biafra is the most expanded.

320 In the PRECIP case (Figure 9c), we can observe that precipitations force
321 really limited SSS variations with a maximum of 2.5 psu. Surprisingly, the most

322 important variability takes place near 3°S between the Bight of Biafra and the
323 Congo plume region, where precipitations are not the strongest (see for example
324 Figure 3 of Da-Allada et al (2013a)). For the RUNOFF case (Figure 9d), some
325 features appear similar to REF. In particular, the variability in the Congo plume
326 region is of the same order, around 8 psu, concentrated along the coast, decreasing
327 rapidly offshore. In the northern part of the basin, around the Bight of Biafra, the
328 variability is present, although lower and not as extended spatially as in REF.

329 Figure 9b shows the sum of the variability of the PRECIP and RUNOFF
330 experiments. The resulting map is similar to REF at first order, capturing the
331 two regions of maximum variability as well as their amplitude. Differences appear
332 though, demonstrating that the full solution cannot be constructed from a linear
333 response to either forcing separately. The variability is larger along the northern
334 coast in the region of the Guinea Current for PRECIP+RUNOFF compared to
335 REF, but lower at the equator and south of it along the African coast. These
336 differences are due to nonlinear effects of the dynamics as well as to the different
337 phases of the seasonal variations of precipitations and runoffs.

338 **4 Mixed layer budget for salinity**

339 4.1 Methodology

340 Following the recent work of Da-Allada et al (2013a) on mixed layer budget for
341 salinity in the Tropical Atlantic using observations, we perform mixed layer salinity
342 budget with our REF experiment. We use the methodology developed by Vialard
343 et al (2001) but applied to the salinity according to the Equation 1 (and using
344 the $\langle . \rangle$ operator defined in Equation 2 for vertical integration) with S the

345 mixed layer salinity, u , v and w the zonal, meridional and vertical velocities, D_l
 346 the horizontal diffusive operator, h the mixed layer depth, E the evaporation, P
 347 the precipitations and R the river runoffs.

$$\begin{aligned} \partial_t S = & - \underbrace{\langle u\partial_x S + v\partial_y S \rangle + \langle D_l(S) \rangle}_A \\ & + \underbrace{w_{z=h}(\bar{S} - S_{z=h}) + \frac{k\partial_z S_{z=h}}{h} - \frac{1}{h} \frac{\partial h}{\partial t} [S_{z=0} - S_{z=h}]}_B \\ & + \underbrace{\frac{1}{h}(E - P - R)SSS}_C \end{aligned} \quad (1)$$

$$\langle x \rangle = \frac{1}{h} \int_0^h x dz \quad (2)$$

348 The terms of the Equation 1 are grouped following Vialard et al (2012): the
 349 vertically averaged horizontal advection and diffusion (A), the vertical advection,
 350 mixing and entrainment (B) and finally the forcing terms : evaporation, precipita-
 351 tions and river runoffs (C). A represents the horizontal transport of salt between
 352 the different regions. As the horizontal diffusion is negligible compared to the hori-
 353 zontal transport (Berger 2012), this term is referenced as "advection" hereafter. B
 354 represents the exchanges between the surface and the subsurface occurring across
 355 the mixed layer. The entrainment, which depicts the mixing effect due to the vari-
 356 ations of the mixed layer depth, is computed as a residual to close the budget at
 357 each time step. The entrainment of Da-Allada et al (2013a) corresponds to our B
 358 term. Finally, C represents the freshwater fluxes across the surface. All these terms
 359 have been evaluated for the two coastal regions with the highest SSS variability
 360 (Figure 9): the Bight of Biafra and the Congo Plume, which appear as black boxes

361 on Figure 1. Budgets are evaluated on-line and archived over successive 5-days
362 periods.

363 4.2 Results

364 4.2.1 The Bight of Biafra

365 Before considering the mixed layer salinity balance, let us describe the seasonal
366 cycle in our model. For this purpose, Figure 10 shows the seasonal cycle of the
367 mixed layer salinity, the salinity below the mixed layer and the mixed layer depth
368 in the model and observations. The bottom panel shows the seasonal cycle of
369 vertical velocity in the model. In this region, the REF experiment underestimates
370 the mixed layer salinity compared to the ARV09 climatology, with a bias reaching
371 2 psu. In addition the model presents a 2 months lag with the climatology when
372 the salinization occurs from May to August (Figure 10a). It is interesting to note
373 that using a different climatology and a different method, Da-Allada et al (2013a)
374 obtain a similar lag in the salinity variations (although they have no bias); it is
375 unclear whether the phase lag comes from similar deficiencies (in forcings or in the
376 climatology) or from independent errors in the two calculations. Apart from these
377 issues, the annual cycle is correctly reproduced in our model, with two maxima.
378 The model also reproduces in a satisfying manner the evolution of the mixed layer,
379 which is very shallow from December to June (12 m on average) when the salinity
380 is low and deepens as the salinity increases from June to September.

381 The vertical velocity (Figure 10c) is positive during most of the year. This may
382 seem surprising, considering that the Ekman transport due to the trade winds
383 generates downwelling north of the equator along the coast of the Gulf of Guinea

384 (see the map of Ekman pumping in Figure 12 of Giordani and Caniaux (2011)).
385 However our model is in agreement with the vertical velocity estimate of Giordani
386 and Caniaux (2011), which is positive in most of the Bight of Biafra region due
387 to the contribution of nonlinear terms. Thus vertical advection, on average, makes
388 the mixed layer saltier. However, vertical advection cannot cause the increase
389 in salinity from May to August, because vertical velocity decreases during that
390 period, due to the increased downwelling tendency because of the intensification
391 of trade winds (Giordani and Caniaux 2011). From May to August, when the
392 mixed layer salinity increases, the mixed layer depth increases and the salinity
393 10 meters below decreases (Figure 10a): this is consistent with vertical mixing. It
394 can be explained by an intensification of the winds, causing larger vertical shears
395 and turbulent fluxes of momentum (Giordani and Caniaux 2011). Note that in
396 the climatology, this link between salinization and mixed layer thickness does not
397 exist, as the deepening of the mixed layer occurs 3 months after the increase of
398 the salinity. Later in the year, from September to December, the model salinity
399 decreases during a time where vertical velocities intensify, which means that the
400 mixed layer shallowing and reduced vertical mixing are the main causes of this
401 freshening.

402 The seasonal cycle of the salinity budget (Equation 1) is shown in Figure 11a.
403 The freshwater fluxes (forcing, green curve) always contribute to diminish the
404 salinity as the evaporation never compensates the precipitations and river runoffs.
405 The forcing does not explain the salinity tendency, whatever the period we are
406 interested in and contrary to the proposals of Dessier and Donguy (1994), Delcroix
407 et al (2005) and Reverdin et al (2007). On the other hand, these results agree with

408 Da-Allada et al (2013a) as they demonstrate the weak influence of the freshwater
409 forcing in the Gulf of Guinea.

410 The most important contributions to the salinity tendency come from the
411 dynamics. Indeed, both the advection and the subsurface contributions are on
412 average higher than the forcing term by an order of magnitude (Figure 11a). They
413 reach their maximum values from May to June with $-1.5 \text{ psu}\cdot\text{month}^{-1}$ on average
414 for the advection and up to $2 \text{ psu}\cdot\text{month}^{-1}$ for subsurface mechanisms. These value
415 are twice the ones of Da-Allada et al (2013a) for the Gulf of Guinea, but it is not
416 surprising as we have higher salinity gradients in a smaller region, increasing the
417 importance of the dynamics.

418 To quantify the contribution of each mechanism and determine which one of
419 them drives the intraseasonal to seasonal tendencies over the 1995-2006 period, we
420 computed monthly linear regression coefficients for advection, subsurface processes
421 and forcing. With $X_t = X_{adv}(t) + X_{sub}(t) + X_f(t)$ the total salinity tendency, equal
422 to the sum of its contributors, the linear regression coefficient α_i of X_i on X_t can
423 be estimated following Equation 3:

$$\alpha_i = \text{cor}(X_i, X) * \text{stdev}(X_i) / \text{stdev}(X). \quad (3)$$

424 In the REF experiment, the horizontal advection drives the salinity tendency
425 variability (Figure 11, bottom panel). Indeed, with a regression coefficient going
426 from 0.75 to 1.5, the advection is the main driver of the total variability. The re-
427 gression coefficient is larger than one from october to march because the variance of
428 the advection is larger than the variance of the total tendency. During that period,
429 on the contrary, the subsurface processes damp the evolution of the salinity in the
430 mixed layer, as shown by the opposite phases of the total and subsurface tenden-

431 cies from September to March (Figure 11, top panel) and the negative regression
432 coefficient (Figure 11, bottom panel). During the salinization period between May
433 and August, the freshening tendency due to horizontal advection progressively de-
434 creases while the subsurface salinization remains always more important, linked
435 with vertical mixing as noted above. Again this is consistent with Da-Allada et al
436 (2013a) who find that entrainment explains the positive tendency of the salinity.
437 Nevertheless, even in May-August, horizontal advection explains more of the ten-
438 dency (the regression coefficient is lower for subsurface processes). This is due to
439 the presence of high frequency variability in the total tendency as well as in the
440 horizontal advection term, while subsurface processes vary on longer time scales.

441 *4.2.2 Congo Plume*

442 In this region, the REF experiment reproduces quite well the evolution of the
443 annual salinity in the mixed layer (Figure 12a) with little bias nor phase shift
444 compared with observations. The salinity increase occurs earlier than in the Bight
445 of Biafra (April-June), followed by weaker variations from July to September.
446 The mixed layer in the model is too shallow compared to the climatology of
447 de Boyer Montegut et al (2004) but it seems more similar to the one of Gior-
448 dani and Caniaux (2011). Like in the Bight of Biafra, the deepening of the mixed
449 layer begins when the salinization occurs from April to August, but it remains
450 limited to 3 meters contrary to the Bight of Biafra where the depth of the mixed
451 layer doubles. On Figure 12a, we can also observe that in the Congo plume region
452 region, the salinities in the mixed layer and 10 meters below it evolve similarly,
453 arguing for lower exchange between surface and subsurface layers. Indeed, the
454 deepening of the mixed layer from May to August (Figure 12b) does not corre-

455 spond to a decreasing subsurface salinity, contrary to the Bight of Biafra. Vertical
456 velocities are almost always positive with a strong semi-annual cycle, very similar
457 to the Bight of Biafra. In the Congo plume however, the increase of vertical veloc-
458 ity from February to June coincides with the initial phase of salinization, which
459 suggests that the strong increase of the mixed layer salinity can be due to a direct
460 transport of salt from the subsurface, consistent with upwelling dynamics as the
461 trade winds intensify (Verstraete 1992).

462 Figure 13 presents the different terms of Equation 1 for the Congo Plume
463 region. As in for the Bight of Biafra, the freshwater forcing is weaker than the
464 other terms, in good agreement with Da-Allada et al (2013a). The salinity tendency
465 results from a balance between horizontal advection (which carries salt away from
466 the region) and the vertical processes that bring salt into the mixed layer. The
467 seasonal cycle of advection and tendency is smoother than in the Bight of Biafra
468 (there is less high frequency variability). Regarding the subsurface processes, time
469 series of the advective and diffusive contributions show that the vertical advection
470 is relatively more important than in the Bight of Biafra (Berger 2012).

471 The regression of horizontal and subsurface processes with the tendency (Fig-
472 ure 13, bottom panel) shows that in the Congo plume region, both processes
473 add up to force the total tendency most of the year. Horizontal advection always
474 contributes positively to the total tendency of the mixed layer salinity. Subsur-
475 face processes damp the tendency (negative regression coefficient) only during the
476 months of September and October, when the vertical velocity weakens and the
477 mixed layer deepens. Subsurface dynamics contribute equally and sometimes more
478 to the variability than the horizontal advection, from May to August. During this
479 period, as the advective tendency goes to zero the subsurface dynamics intensify

480 and transport more and more salt to the mixed layer from the subsurface, causing
 481 the strong salinization.

482 4.3 Contribution of transient dynamics to horizontal advection

483 Comparing the salinity tendency on Figures 11 and 13, we find a higher variability
 484 in the Bight of Biafra. This leads us to suspect that the correlation of transient fluc-
 485 tuations of velocity and salinity may be an important contribution to the budget.
 486 To assess this contribution, we decompose the salinity tendency due to horizontal
 487 advection (here $\partial_t S_{adv}$) in two terms :

$$\partial_t S_{adv} = \langle \partial_t S \rangle_{month} + \partial_t S_{res} \quad (4)$$

488 In this equation, $\partial_t \langle S \rangle_{month}$ represents the part of the horizontal advection
 489 due to the seasonal mean velocity and seasonal mean salinity. To compute it, we
 490 first apply a low pass filter on the mean seasonal cycle of the zonal and meridional
 491 velocities as well as the mixed layer salinity to remove all the variability at higher
 492 frequencies than a month. This smoothed seasonal cycle is noted $\langle \rangle_{month}$. The
 493 seasonal advection is thus determined off-line with :

$$\langle \partial_t S \rangle_{month} = - \langle U \rangle_{month} \partial_x \langle S \rangle_{month} - \langle V \rangle_{month} \partial_y \langle S \rangle_{month}$$

494 The $\partial_t S_{res}$ represents the residual, with all the contributions other than the mean
 495 seasonal velocities and salinities: high frequency waves, eddies... It results from the
 496 difference between the on-line budget for the advection (A term of the Equation 1)
 497 and the off-line budget.

498 The results of these computations can be seen on Figure 14. In the Bight of
 499 Biafra, the residual (eddy) term dominates the horizontal advection tendency while

500 in the Congo Plume the total advection is almost entirely due to the mean seasonal
501 cycle of velocity and salinity, underlying the dynamical differences between the
502 two regions. This agrees with the model results of Guiavarc'h et al (2009) who
503 point out that the surface Eddy Kinetic Energy (EKE) at periods between 10
504 and 20 days is much higher in the Bight of Biafra than in the Congo Plume.
505 In their model, the surface intensification of EKE is due to the variability of
506 the winds which is higher north of the equator than to the south. This surface
507 intensification of EKE is validated at one location by current meter measurements
508 that indicate an even higher surface EKE than the model (see Guiavarc'h et al
509 (2009) Figure 7). Another reason for the high residual in Figure 14a is that our
510 Bight of Biafra region encompasses the equator, where the 10-20 days variability is
511 especially large. Offshore Angola and Gabon, the north-westward surface currents
512 are spatially smooth and do not vary as much in direction, being the coastal
513 part of the wind forced South Equatorial Current (Stramma and Schott 1999).
514 As a consequence, the horizontal advection is mostly due to the mean seasonal
515 component.

516 **5 Conclusions**

517 In this paper, we analyse the mechanisms of the intra-seasonal and seasonal vari-
518 ability of the SSS in the eastern Gulf of Guinea. We evaluate the importance
519 of the different sources of freshwater using numerical modelling and we quantify
520 the dynamical contribution to the SSS variability using mixed layer budgets for
521 salinity.

522 Our sensitivity experiments, forced by either runoffs or precipitations, empha-
523 sise the causal relationship between the water flux forcing and the SSS in the
524 eastern Gulf of Guinea. It appears that the river runoffs, despite the fact that
525 their volume represents only 38% of the total freshwater inflow, are necessary to
526 explain the amplitude of the seasonal cycle of the SSS (which reaches 6 psu or
527 more along the coast) while precipitations alone generate a weaker seasonal cycle
528 with an amplitude of about 2 psu. However, these sensitivity studies, carried out
529 with a fully nonlinear model, do not imply that there is a simple local relationship
530 between the freshwater forcing and the SSS in a given region. Indeed, in a recent
531 study based on satellite observations, Tzortzi et al (2013) failed to establish such a
532 relationship for the Gulf of Guinea and suggested that advection and mixing must
533 play a role locally to explain the spatial structure and the phase of the seasonal
534 cycle.

535 Mixed layer budgets in the Bight of Biafra and the Congo plume demonstrate
536 the importance of the dynamics, in good agreement with the recent findings of
537 Da-Allada et al (2013a) and the hypothesis of Tzortzi et al (2013). In both the
538 Bight of Biafra and the Congo plume region, the surface circulation is responsible
539 for an offshore transport of coastal freshwater and thus tends to decrease the
540 mixed layer salinity. The intensity of this transport depends on the amount of the
541 river discharge and is logically minimum between June and August (Figures 7, 11
542 and 13). Thus the freshwater input appears as a limiting factor for horizontal
543 advection, even though it does not drive directly the variability of the salt content
544 in the mixed layer. The vertical physics, which are responsible for the salinization
545 from May to August when the horizontal advection weakens, differ in each region.
546 In the Bight of Biafra, the salt transport from the subsurface layers comes from

547 an intensification of the mixing. In the Congo plume, the upwelling dynamics
548 dominate the salinization, vertical advection representing the main contribution
549 to the subsurface salt input. Overall, horizontal advection is the main driver of SSS
550 variability in the Bight of Biafra, while vertical processes damp the variability, as
551 demonstrated by a regression analysis. The picture is more complex for the Congo
552 plume region, with horizontal advection and vertical processes both contributing
553 positively to the SSS tendency during most of the year.

554 Finally, we have calculated separately the advective contributions due to the
555 mean seasonal cycle of horizontal velocity and salinity. We have shown that this
556 seasonal contribution explains the advection in the Congo plume. On the contrary,
557 transient dynamics such as high frequency waves or eddies dominate the horizontal
558 advection in the Bight of Biafra. The near equatorial position of the enclosed
559 Bight of Biafra may explain this difference. A specific study of the variability
560 and its contrast between the northern and southern part of the bight would be
561 interesting, but it may require a higher resolution model.

562 Although our mixed layer budget agrees overall with the observation-based
563 estimate of Da-Allada et al (2013a), there are differences in the strength of the
564 different terms as well as in the details of the seasonal cycle. For example, the semi
565 annual cycle of the SSS is much more pronounced in our model, which may be due
566 to different forcings (we use precipitations from Large and Yeager (2009), while Da-
567 Allada et al (2013a) use precipitations from ERA-Interim). Recently, Da-Allada
568 et al (2013b) have used the numerical model of Jouanno et al (2011) in order to
569 study the inter-annual variability of salinity in the Gulf of Guinea. The seasonal
570 cycle of their model presents differences with ours that will need to be investigated.
571 For example, their SSS is closer to the climatology than ours, which may be due

572 to different choices for the forcing or the vertical mixing parameterization. Despite
573 the differences in the mean state of the two models, our main results regarding
574 the seasonal budgets for the mixed layer salinity are consistent, which gives us
575 confidence that they are robust. More in-situ and satellite observations are clearly
576 needed to conduct more in-depth validations of these numerical models, at the
577 process level. In the Congo plume region, the PIRATA mooring at 6°S-8°E, now
578 operational again, will provide extremely valuable long time series. Similar long-
579 term observations are crucially needed in the Bight of Biafra.

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698 845

Table 1 Mean volume of freshwater discharge in the different regions of the Gulf of Guinea. The regions are the ones described in Figure 1. Precipitations come from the DFS4.3 product and river runoffs from Dai et al (2009).

Regions	Precipitations (Sv)			Coastal Runoffs (Sv)		
	mean	max	min	mean	max	min
Gulf of Guinea	0.135	0.258	0.058	0.082	0.099	0.066
Bight of Biafra	0.027	0.040	0.014	0.027	0.039	0.020
Congo region	0.014	0.031	0.0001	0.044	0.060	0.033

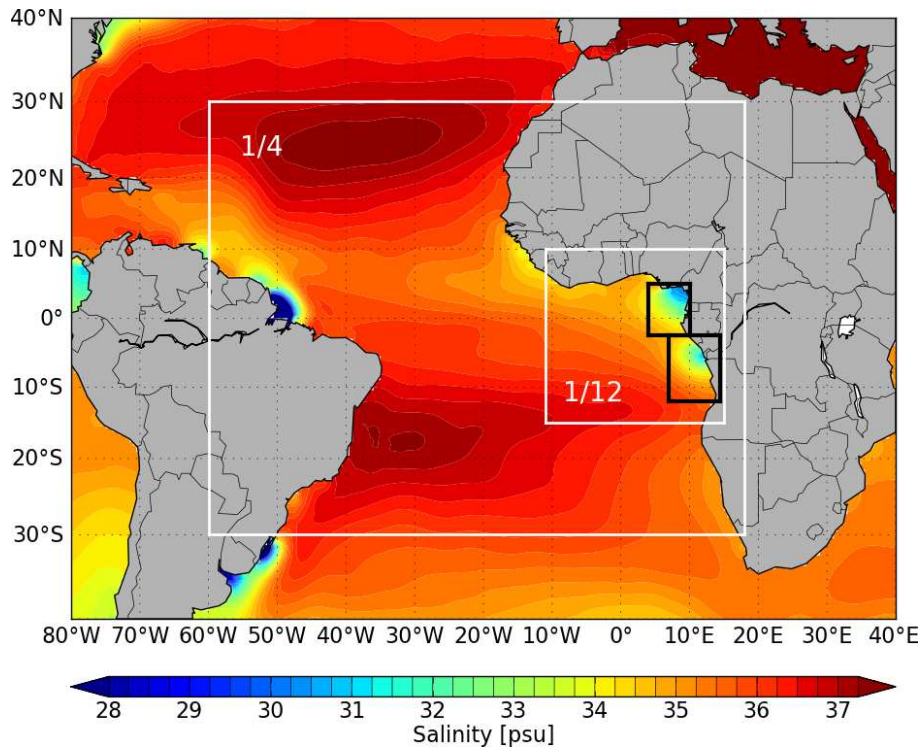


Fig. 1 Mean SSS for the tropical Atlantic from ARV09 climatology (Gaillard et al 2009). The models domains with $1/12^\circ$ and $1/4^\circ$ resolution used for this study are outlined in white, and the domains used for freshwater impacts and mixed layer budget analysis are indicated in black (Biafra box, north of the equator, and Congo box, south of the equator)

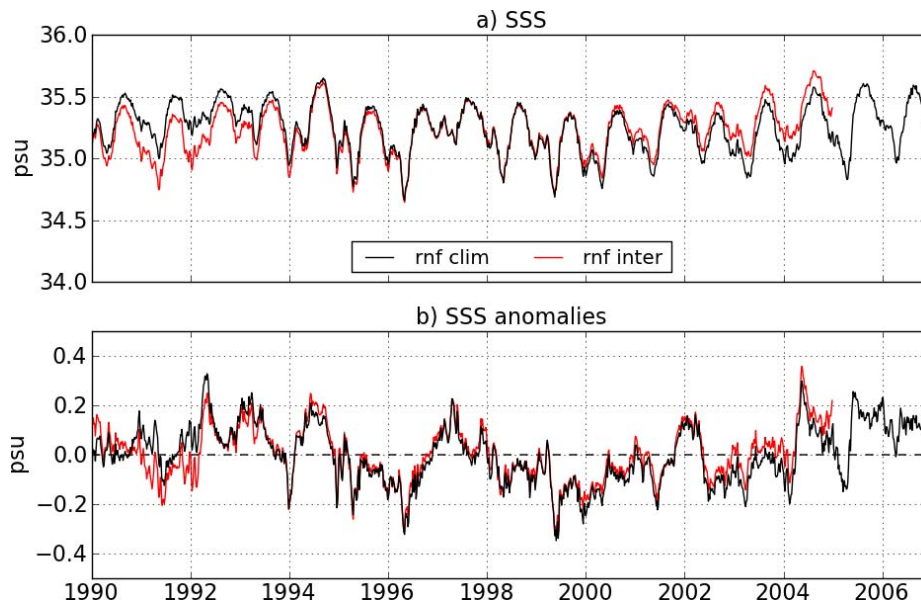


Fig. 2 a : Mean SSS of the model in the Gulf of Guinea ($1/12^\circ$ domain of Figure 1) using climatological (black) and inter-annual (red) runoffs from 1990 to 2006 (2004 for inter-annual). b: corresponding anomalies for climatological (black) and inter-annual (red) runoffs. The anomalies are computed using normalized time series where the long term trend has been removed. Climatological runoffs have been computed by averaging the inter-annual runoffs data of Dai et al (2009) from 1990 to 2004.

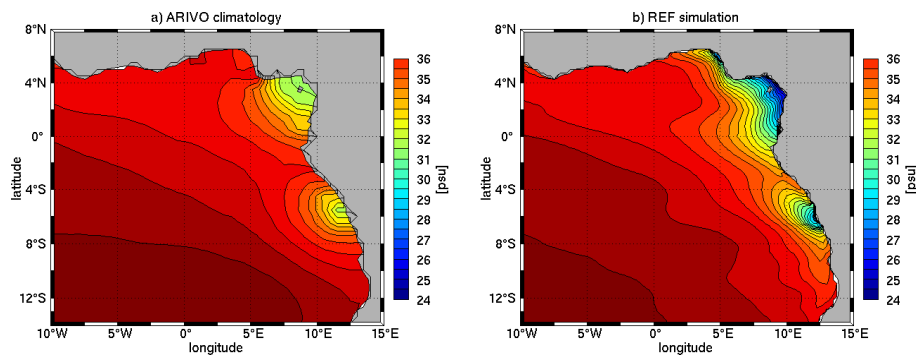


Fig. 3 Mean state of SSS in the Gulf of Guinea in the ARV0 climatology (Gaillard et al 2009) (a) and in the REF experiment from 1995 to 2006 (b).

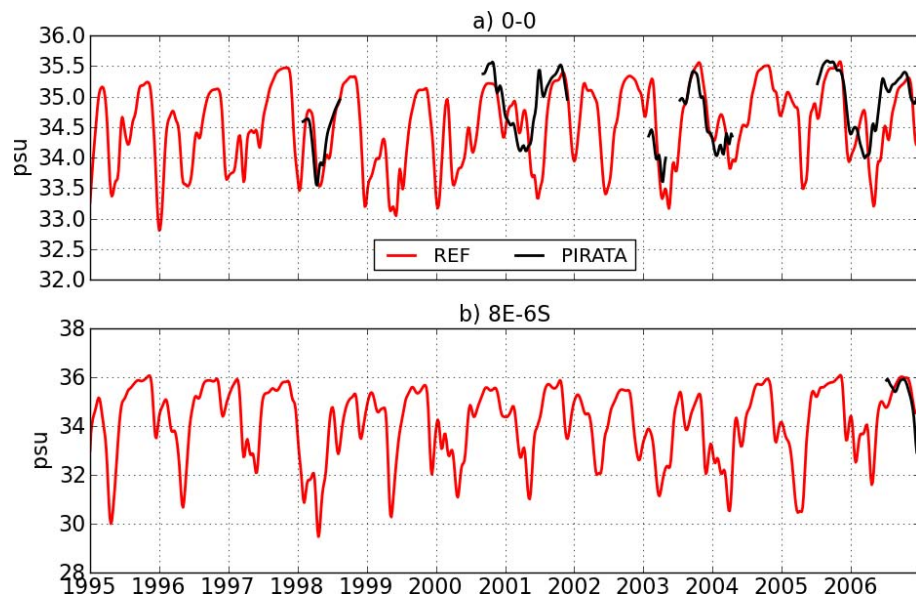


Fig. 4 Inter-annual series of the SSS on the positions of two PIRATA moorings in the Gulf of Guinea: at $0^{\circ} 0^{\circ}$ (a) and $8^{\circ}\text{E } 6^{\circ}\text{S}$ (b) for the REF experiment (red) and the PIRATA moorings observations (black) from 1995 to 2006.

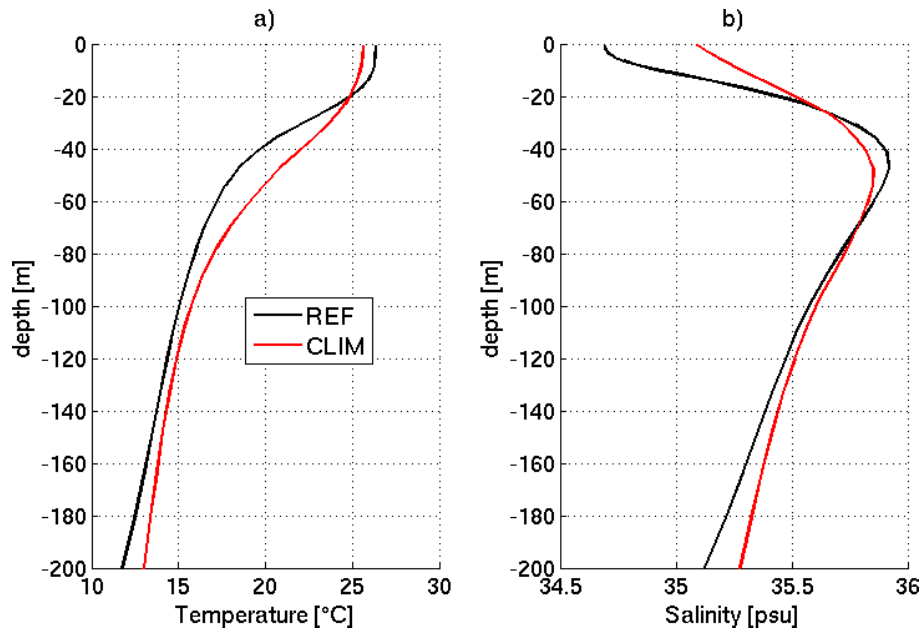


Fig. 5 Mean temperature (a) and salinity (b) profiles in the Gulf of Guinea from the surface to 200 meters in the ARV09 climatology (red) and the REF experiment from 1995 to 2006 (black). The domain used correspond to the $1/12^\circ$ domain visible on Figure 1.

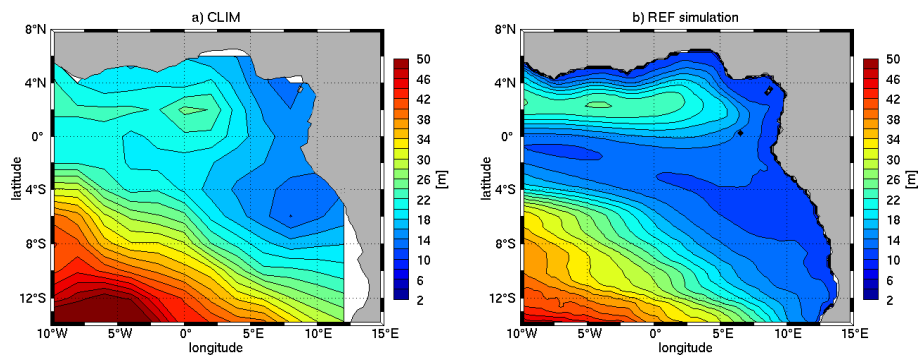


Fig. 6 Mean state of the mixed layer depth in the Gulf of Guinea in the climatology of de Boyer Montegut et al (2004) (a) and in the REF experiment from 1995 to 2006 (b).

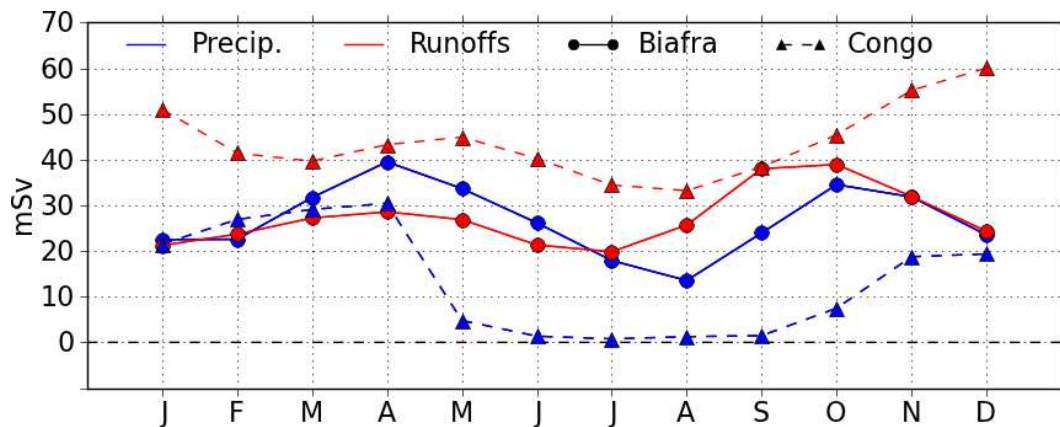


Fig. 7 Mean seasonal cycle of the freshwater input in the Biafra and Congo boxes (Figure 1).

Precipitations and runoffs for each region appear as blue and red lines respectively. Data for the Bight of Biafra are marked by continuous lines with circles and data for the Congo plume are marked by dashed lines with triangles. Precipitations (GPCP) come from the Drakkar Forcing Set (Brodeau et al 2010) and the river runoffs from Dai et al (2009).

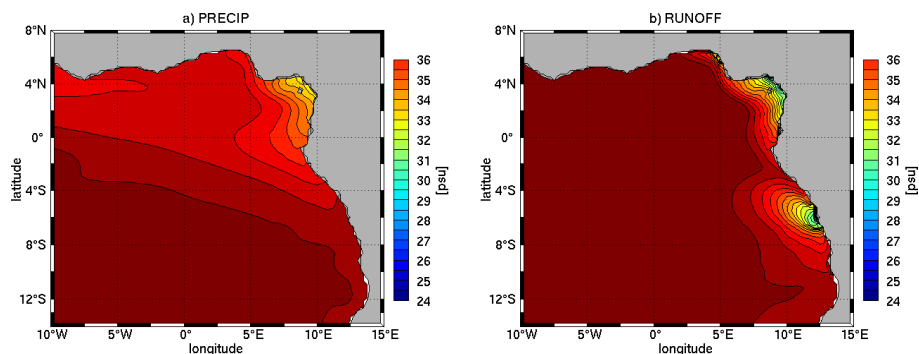


Fig. 8 Mean state of the SSS in the Gulf of Guinea in our PRECIP (a) and RUNOFF (b)

sensitivity experiments from 1995 to 2006.

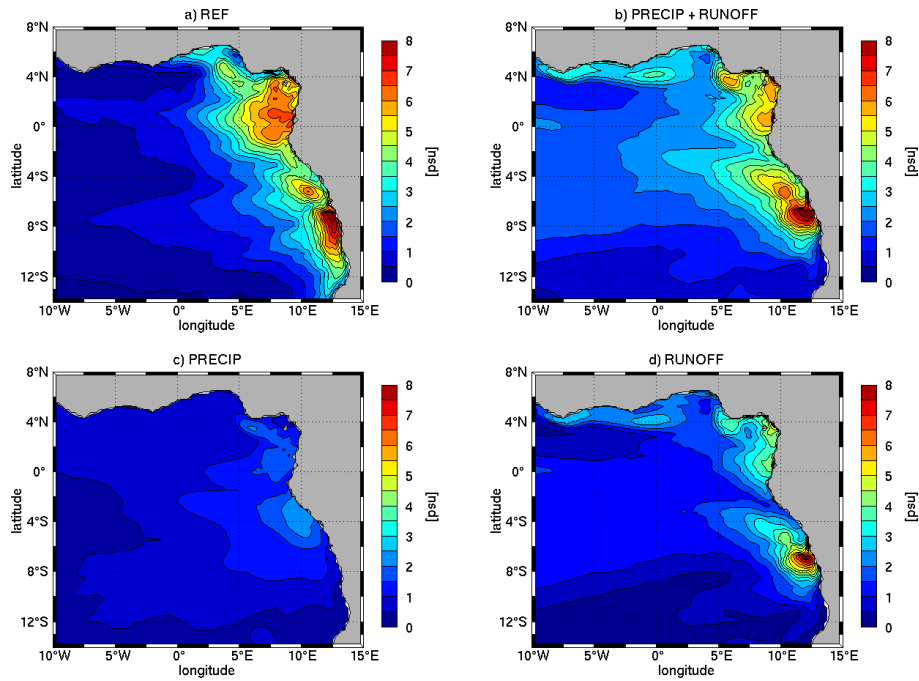


Fig. 9 Map of the SSS seasonal variability amplitude in the Gulf of Guinea. a) REF experiment, b) sum of the variability from PRECIP and RUNOFF experiments, c) PRECIP experiment, d) RUNOFF experiment. The seasonal amplitudes are computed at each grid point from a time series of monthly SSS from 1995 to 2006, by taking the difference between the maximum and the minimum monthly SSS for each year, and then averaging these amplitudes for all years.

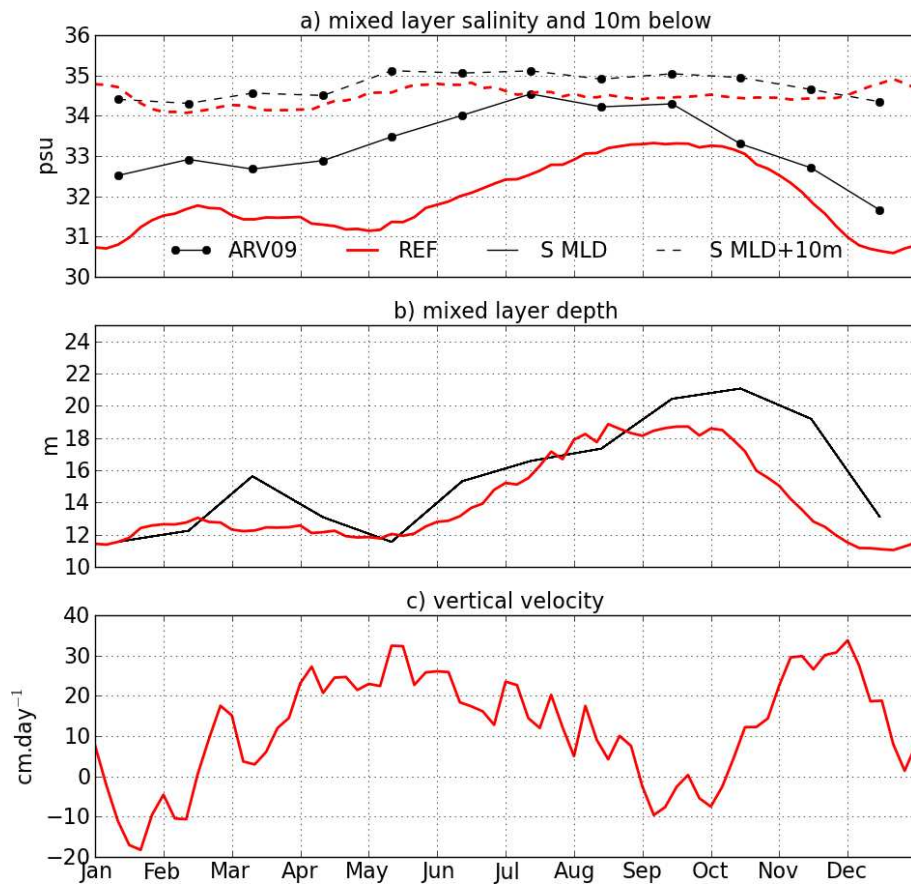


Fig. 10 Mean seasonal values of various fields computed from 1995 to 2006 in the Biafra box of Figure 1. a) Mean annual salinity in the mixed layer (continuous) and 10 meters below (dashed) in REF (red) and ARV09 (black). b) Mean annual evolution of the mixed layer depth in REF (red) and the climatology of de Boyer Montegut et al (2004) (black). c) Mean annual vertical velocity in the REF experiment.

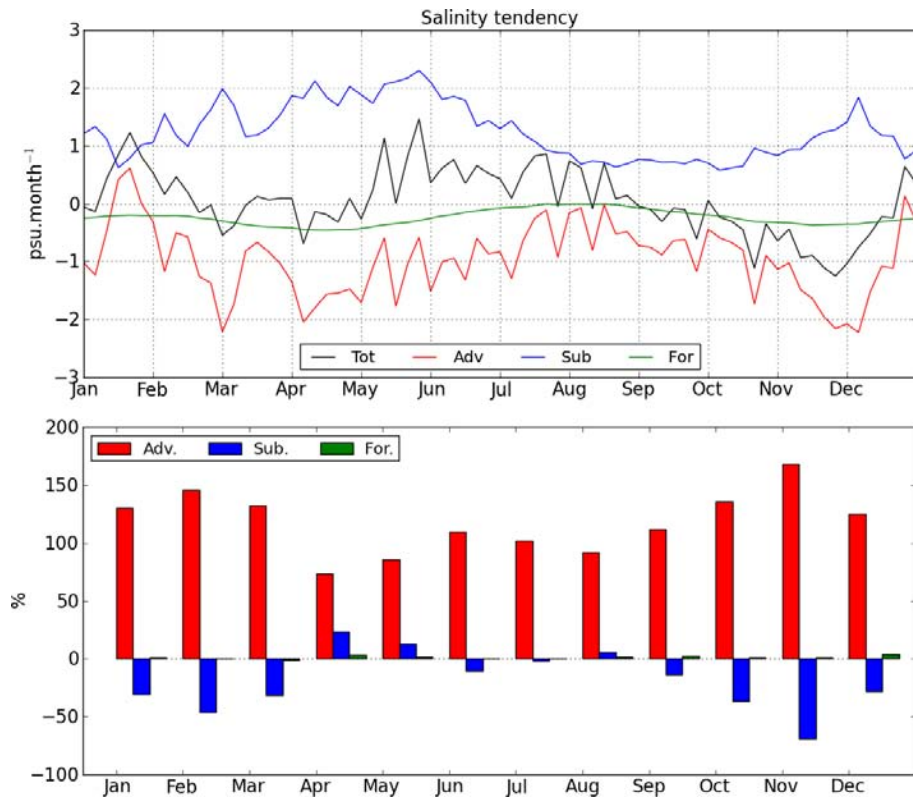


Fig. 11 Top : mean seasonal contributions to the mixed layer budget for salinity of the A, B and C terms of Equation 1. These contributions have been computed from 1995 to 2006 in the Biafra box of Figure 1. Bottom : monthly regression coefficients of the terms of Equation 1 computed using the Equation 3. On these figures, the total trend appears in black, the advection in red, the subsurface processes in blue and the forcing in green.

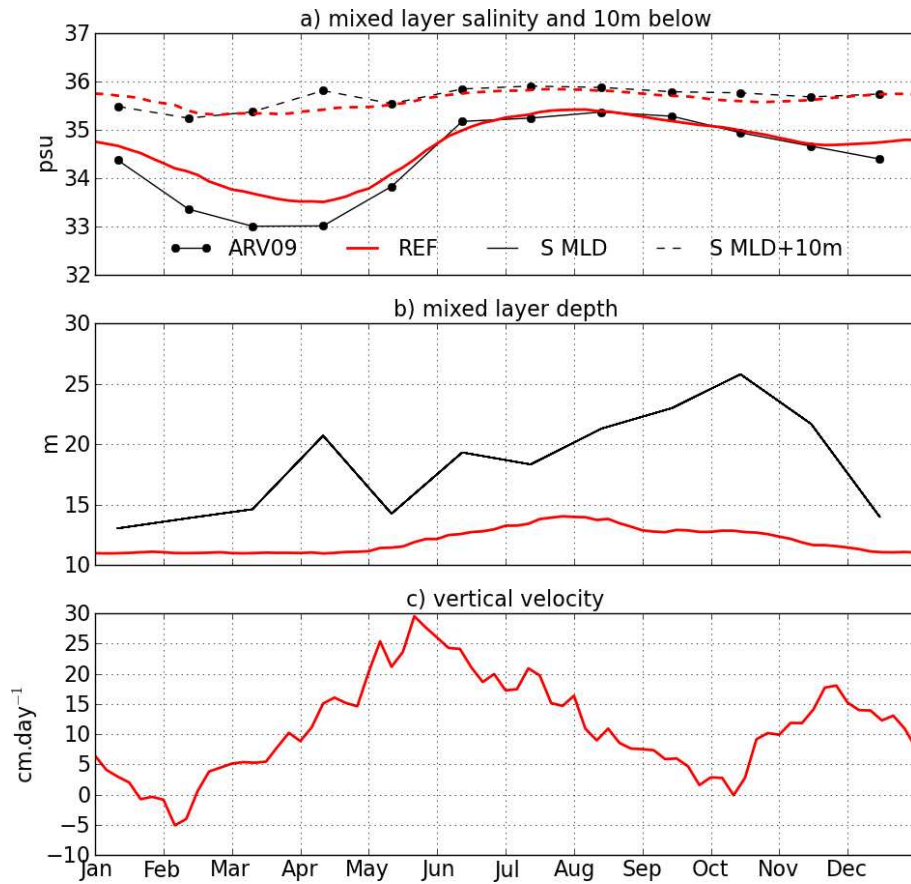


Fig. 12 Mean seasonal values of various fields computed from 1995 to 2006 in the Congo box of Figure 1. a) Mean annual salinity in the mixed layer (continuous) and 10 meters below (dashed) in REF (red) and ARV09 (black). b) Mean annual evolution of the mixed layer depth in REF (red) and the climatology of de Boyer Montegut et al (2004) (black). c) Mean annual vertical velocity in the REF experiment.

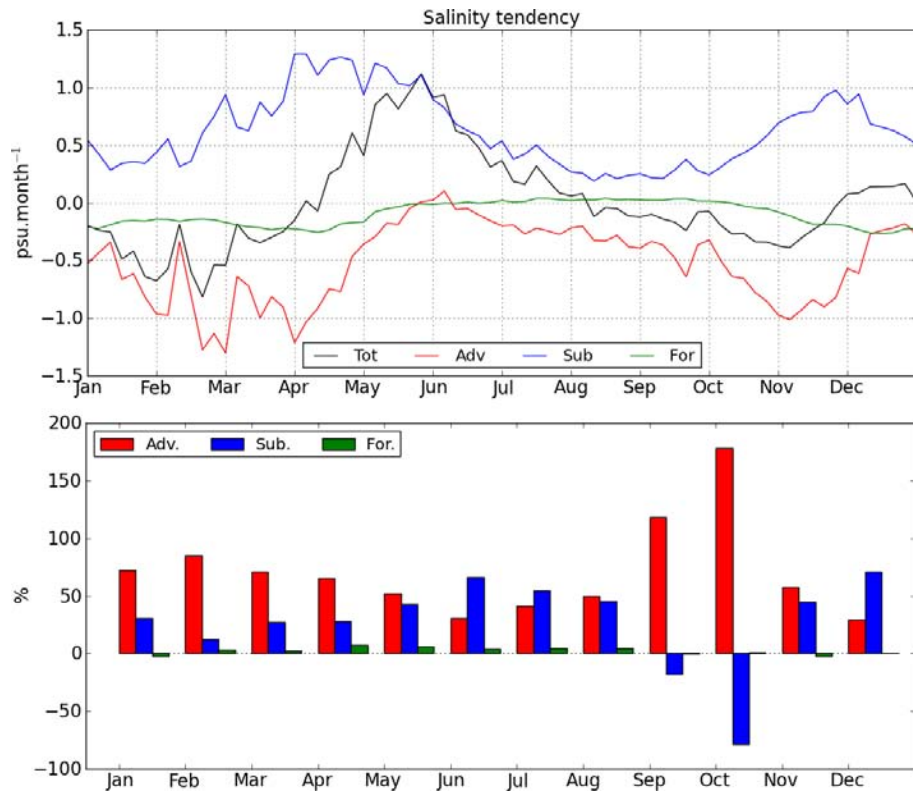


Fig. 13 Top : mean seasonal contributions to the mixed layer budget for salinity of the A, B and C terms of Equation 1. These contributions have been computed from 1995 to 2006 in the Congo box of Figure 1. Bottom : monthly regression coefficients of the terms of Equation 1 computed using the Equation 3. On these figures, the total trend appears in black, the advection in red, the subsurface processes in blue and the forcing in green.

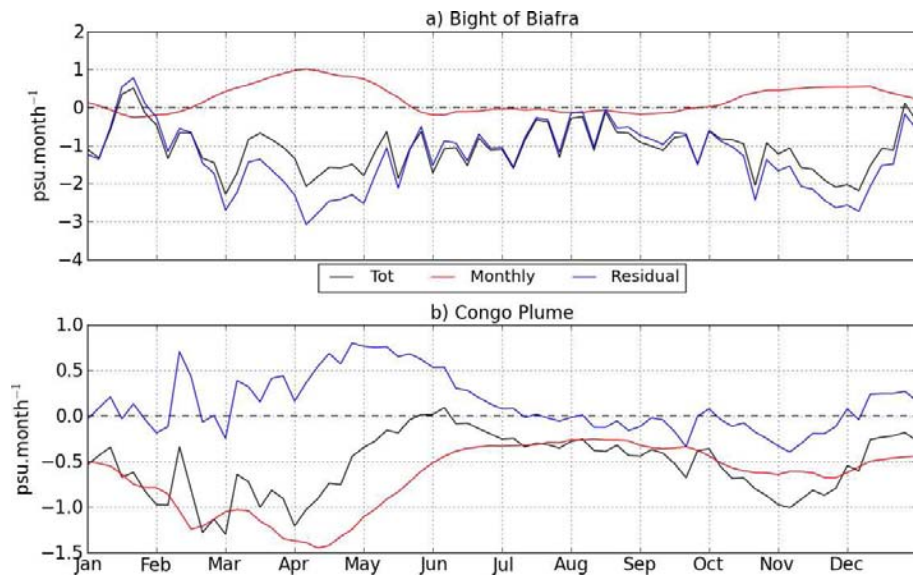


Fig. 14 Decomposition of the horizontal advective term of Equation 1 (in black) into a contribution from the monthly mean seasonal cycle (red) and a residual due to other variabilities such as eddies and high frequency waves (blue) (see Equation 4).