Clear-Sky Turbulence and Shallow Convection: New Insights Combining SAR Images, Satellite Brightness Temperature and In Situ Measurements

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Abstract

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Key Points:

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15	•	Atmospheric coherent structures (cells, rolls and cold pools) are systematically de-
16		tected and analysed in a high-resolution SAR image
17	•	Properties of rolls from SAR measurements are comparable with the turbulence
18		organization deduced from airborne data

 A diversity of cold pool geometrical and dynamical features is related to cloud intensity provided by satellite brightness temperature

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21 Abstract

The imprint of marine atmospheric boundary layer (MABL) dynamical structures on sea 22 surface roughness, as seen from Sentinel-1 Synthetic Aperture Radar (SAR) acquisitions, 23 is investigated. We focus on February 13th, 2020, a case study of the EUREC4A (Elu-24 cidating the role of clouds-circulation coupling in climate) field campaign. For clear sky 25 conditions, convective rolls and cells imprints on sea surface roughness is confirmed through 26 the intercomparison with MABL turbulent organization deduced from airborne measure-27 ments. A discretization of the SAR wide swath into $25 \ge 25 \text{ km}^2$ tiles allows us to cap-28 ture the spatial variability of the turbulence organization varying from rolls to cells. We 29 then objectively detect cold pools within the SAR image and combine them with geo-30 stationary brightness temperature. The geometrical or physically-based metrics of cold 31 pools are correlated to cloud properties. This provides a promising methodology to an-32 alyze the dynamics of convective systems as seen from below and above. 33

³⁴ Plain Language Summary

We propose an innovative approach to investigate the marine atmospheric bound-35 ary layer (MABL) dynamical structures by combining spaceborne Synthetic Aperture 36 Radar (SAR) images, brightness temperature (T_B) from the Geostationary Operational 37 Environmental Satellite (GOES) and in situ turbulence airborne measurements from the 38 EUREC4A field campaign. Focusing on February 13th, 2020, two types of atmospheric 39 processes are investigated: clear sky organizations and cold pools. The signature of at-40 mospheric coherent structures on sea surface roughness, especially convective rolls, is val-41 idated with respect to the turbulence measurements of the ATR 42 aircraft. The cold 42 pools are detected within the SAR image using a segmentation method. Cold pool char-43 acteristics such the size and the gust front intensity can then be directly derived from 44 the SAR image. The GOES images provide cloud field properties with the temporal di-45 mension. Exploring backward cloud evolution with respect to the SAR image timing ap-46 pears meaningful to catch the life cycle of cold pools and convective clouds from which 47 they originate. The application of this approach could pave the way to access the dy-48 namics of convective systems as seen from below and above, allowing to go one step fur-49 ther in the quantitative use of SAR images to investigate boundary layer processes. 50

51 **1** Introduction

Marine atmospheric boundary layer (MABL) dynamics plays a crucial role in the 52 mesoscale organization of convection. Among the different kinds of coherent organiza-53 tions occurring inside the MABL, three-dimensional cells and quasi-two-dimensional con-54 vective rolls are frequent (Etling & Brown, 1993; Weckwerth et al., 1996; Atkinson & Wu Zhang, 55 1996; Wang et al., 2020). The roll wavelengths are in the range of spectral scales that 56 contribute the most to turbulence kinetic energy (Lemone, 1976). It leads to a substan-57 tial impact on the vertical transport of heat and moisture (Chou & Ferguson, 1991; Etling 58 & Brown, 1993; Brilouet et al., 2020), which is still not completely understood. Another 59 important kind of MABL coherent structures are the cold pools which are key compo-60 nents for the life cycle of convection and cloud organization. Cold pools are generated 61 by the downward motion of air mass cooled by rain drop evaporation. At the surface, 62 it spreads out horizontally to form a gust front with an enhanced wind intensity. Pre-63 vious studies have highlighted the relationships between cold pools, their accompany-64 ing moisture distribution and related cloud cover (Khairoutdinov & Randall, 2006; Zuidema 65 et al., 2012; Torri et al., 2015). Cold pools can, in their core, inhibit the development 66 of new convective cells due to their stabilizing effects. However, they can also dynam-67 ically trigger, at their edges, new convective cells, through wind convergence and lifting 68 or thermodynamically by accumulation of moisture (Lima & Wilson, 2008; Feng et al., 69

2015; Torri & Kuang, 2019). It is then fundamental to improve the monitoring of cold
 pools and their interplay with convection systems.

To improve understanding of MABL coherent structures, accurate measurements 72 are needed. Most observational studies are based on in-situ data, including shipboard 73 (e.g., Zuidema et al., 2012; de Szoeke et al., 2017) or shore-based measurements (e.g., 74 Vogel et al., 2021): those provide thermodynamical and dynamical measurements, with 75 however spatially-limited sampling. For shallow convection MABL coherent structures, 76 spatial sampling could be obtained with high-resolution (< 100 m) spaceborne obser-77 78 vations of sea-surface roughness, related to surface wind and stability variations. In particular, SAR backscatter is such a high resolution ($\sim 10 \text{ m} - 100 \text{ m}$) measurement, avail-79 able during day and night, regardless of weather conditions and cloud cover. Coherent 80 structures induce a contrasts of SAR backscatter, which are however difficult to extract 81 since they are multiscale and superimpose with other (oceanic) signatures (Kudryavtsev 82 et al., 2005; Avet et al., 2021). Atlas (1994) was a pioneer in highlighting the potential 83 signature of a storm microburst on the sea surface. Further work focused on convective 84 rolls (Alpers & Brümmer, 1994; Young et al., 2000; Vandemark et al., 2001; Wang et al., 85 2020), their link with stratification Stopa et al. (2022) and on deep convection cold pools 86 from SAR winds (La et al., 2020) or from scatterometer winds (Mapes et al., 2009; Kil-87 patrick & Xie, 2015; Garg et al., 2020). 88

Systematic and quantitative extraction of high resolution information on MABL 89 properties in SAR measurements is still a challenge, due to overlapping geophysical pro-90 cesses which require complementary information. The aim of this paper is thus to present 91 an innovative approach based on a combined use of SAR and geostationary satellite data 92 to study the interplay between the MABL and convective organization. By doing so we 93 (i) show how SAR clear-sky convection compares to in situ measurements and (ii) anal-94 yse the morphological and dynamical properties of cold pools, related to moist convec-95 tion. 96

The paper is organized as follows: the data and methods are presented in Section
Section 3 is devoted to the clear-sky turbulence and Section 4 concerns the cold pool detection and characterization. The last section concludes the paper with a discussion of the main results.

¹⁰¹ 2 Data and Methods

2.1 Data

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We used C-band SAR images from the Sentinel-1 satellite, which probes the sea surface through clouds, in interferometric wide swath (IW) mode. Each image is 400km wide with effective resolution of 100 m, allowing multiscale sampling of cold pools. The C-band backscatter signal (σ_0) is sensitive to the slope distribution of centimetrescale wind-wave, with short adjustment timescales to changes in surface wind (Kudryavtsev et al., 2005). Backscatter contrasts can thus be related to surface signatures of MABL processes (Ayet et al., 2021).

To obtain information on the cloud layer, we used infrared brightness temperature (T_B) provided by the Advanced Baseline Imager (ABI, Schmit et al., 2005) onboard GOES-16 (Geostationary Operational Environmental Satellite). In the region of interest, the Atlantic trades, cloud cover mainly consists of low-level clouds which facilitates the interpretation of the satellite images can be detected with the brightness temperature despite their small temperature differences with the sea surface. The spatial resolution is 2 km and temporal resolution 10 min. In addition, we used wind speed and direction estimated from WindSat, a polarimetric microwave radiometer operated on the Coriolis satellite, with a spatial resolution of 25 km.

Finally, we used in situ observations from the EUREC4A campaign (Elucidating the role of clouds-circulation coupling in climate), which took place over the Western part of the Tropical Atlantic near Barbados, in Jan-Feb 2020 (Stevens et al., 2021). In particular, the SAFIRE ATR 42 aircraft samples the atmosphere around the cloud-base level and at different heights in the subcloud layer (Bony et al., 2022). Among the collected data, we will focus on the vertical velocity turbulent fluctuations (w') deduced from a five-hole radome nose with a spatial resolution of around 4 m (Brilouet et al., 2021).

¹²⁷ Here, we focused on February 13th 2020, with the best spatial and temporal over-¹²⁸ lapping between ATR 42 and Sentinel-1 acquisition: an ATR 42 flight between 0735 UTC ¹²⁹ and 1152 UTC (take-off and landing times) and a Sentinel-1 crossing around 0935 UTC, ¹³⁰ about 100 km away from the aircraft track (see Fig. 1a). Superimposing T_B and σ_0 shows ¹³¹ distinct atmospheric processes: clear-sky convective rolls and cold pools below clouds.

132 **2.2 Methods**

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2.2.1 Characteristics from two-dimensional autocorrelation

Because turbulent processes are highly variable in time and space, as well as very 134 sensitive to their dynamical environment, the wide swath SAR image has been split into 135 $25 \times 25 \text{ km}^2$ tiles (Fig. 2a). This is a good compromise with sub-domains large enough 136 to properly sample coherent structures and small enough to be homogeneous. This is con-137 sistent with previous studies based on 20×20 km² Wave Mode SAR acquisitions which 138 focused on convective roll analysis (Wang et al., 2019; Granero-Belinchon et al., 2022). 139 Using the GOES image, tiles with cloudy pixels were removed from the clear-sky tur-140 bulence analysis (blue tiles in Fig. 2a). Based on in situ EUREC4A data, a T_B thresh-141 old of 292 K has been chosen : colder pixels are considered as cloudy. 142

Two-dimensional autocorrelation is calculated over each 25×25 km² tiled area. 143 Following the methodology detailed in Granero-Belinchon et al. (2022), we first estimate 144 the longitudinal axis of the coherent structures (noted hereafter as Ψ , Fig. 2b,c solid red 145 line). Then, following Lohou et al. (2000), the integral lengthscale L_E is estimated for 146 each angle θ and an elliptical fit is applied to the resulting polar curve (red ellipse in Figs. 2b,c). 147 The type of organization is then diagnosed from the flatness parameter f of the ellipse. 148 A threshold is empirically chosen to distinguish rolls ($f \ge 0.7$, green tiles in Fig. 2a 149 and Fig. 2c) from the transition between rolls and cells (f < 0.7, orange tiles in Fig. 2a 150 and Fig. 2b). When L_E anisotropy is strong enough, the orientation of the major axis 151 provides a second estimate of the roll direction (noted hereafter as α , dashed purple line 152 in Figs. 2b,c). Moreover, for roll tiles, the organized structure lengthscale L_{OS} is defined 153 as the autocorrelation secondary local maximum along the direction perpendicular to 154 the structures (blue line in Fig.2c): it corresponds to the mean transverse lengthscale 155 of the rolls. 156

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2.2.2 Identification and characterization of cold pools

To analyse the morphological and dynamical characteristics of cold pool signatures 158 on SAR, we developed an objective identification method, validated over the case study. 159 An example a cold pool signature is given in Fig. 3a. The method makes the physical 160 161 assumption that the pattern of a cold pool consists of strong positive gradients at its forward and backward edges (blue zones in Fig. 3c) and a rather smooth central area with 162 a negative gradient (green zone in Fig. 3c). This corresponds to the gust front and to 163 the horizontal wind divergence associated to the downdraft, respectively. The edges are 164 first detected by computing increments on the gaussian-filtered backscatter signal and 165

selecting pixels with increments larger than twice the increment standard deviation. Individual cold pools objects (orange contours in Fig. 3b) are then defined by grouping edges
together if a zone of negative increment exists between them (e.g., Fig. 3c, the blue edges
are part of the same cold pool because of the green zone with negative increment). For
the SAR image studied here, this objective methodology captures most of the cold pools,
as validated visually (Fig. 4a). Future work will focus on extending this method systematically to other SAR images.

Detected cold pools were contextualized with an object identification of cloud structures within GOES T_B . The method is described in Brient et al. (2019) and Villefranque et al. (2020): it uses the watershed algorithm in space and time (3D) with two thresholds on the cloud object ($T_B \leq 292$ K) and on the cold cores ($T_B \leq 285$ K), allowing to follow the identified object in time (contours in Fig. 3e). This method thus connects the instantaneous snapshot of cold pools with the temporal evolution of the cloud organization and individual cloud life cycle.

¹⁸⁰ 3 Clear-Sky Turbulence

The ability of SAR to retrieve MABL turbulence activity is assessed using airborne 181 observations. In spite of a non perfect overlap between airborne and spaceborne obser-182 vations, two transects were made in the SAR image to mimic the aircraft legs (Fig. 1a). 183 We assume the stationarity of the turbulent field and an advection with the mean wind 184 direction of the coherent structures contained in the SAR image over the ATR 42 area. 185 The direction of the two SAR transect blue and green lines) was chosen to have the same 186 orientation with respect to WindSat wind than the ATR42 (brown and orange line). The 187 time series of w' measured by the ATR 42 and σ_0 fluctuations highlight similar signal 188 dynamics (Fig. 1b). For the two sampling directions, signals differ significantly: the pe-189 riodicity of the sampled eddies is significantly smaller for the transversal legs (blue and 190 brown) than longitudinal ones (green and orange). This emphasizes the nonaxisymmet-191 ric behavior of both turbulence and roughness fields. 192

To document the uncertainties associated to the unidirectional nature of the in situ 193 airborne sampling, a set of 40 1D-SAR transects is extracted (contained within the pink 194 parallelepipede in Fig. 1a). The L_{OS} computed on this set show a substantial variabil-195 ity between 1.10 km and 4.71 km with an average value of 2.65 km (see the boxplot in 196 Fig. 1f). The estimates made on the ATR legs at 300 m and 600 m are 1.60 km and 2.00 m197 km, respectively, are well within this range of values covered by the 1D-SAR estimates. 198 This result is in line with the study of Vandemark et al. (2001) and consolidates the hy-199 pothesis of a sea surface roughness imprint from MABL coherent structures. One of the 200 SAR strength is to provide a 2D-field of roughness which allows for 2D autocorrelation 201 analysis. In the present case, the estimate of L_{OS} over the pink area from the 2D au-202 tocorrelation, 2.90 km, is well consistent with the 1D SAR range of values and with in 203 situ estimates. In the SAR subdomain considered here, a dropsonde launched from the 204 HALO aircraft close to the SAR passage timing reveals a well-mixed MABL with a depth 205 of $z_i = 570$ m. It leads to an aspect ratio of 5.1. For ATR42 measurements, the MABL 206 depth estimates are variable, ranging from $z_i = 500$ m to $z_i = 800$ m. This variabil-207 ity can be driven by dry air intrusions from the troposphere or by secondary circulation 208 dynamics generated by the mesoscale cloud system. Therefore, the aspect ratio estimates 209 in the ATR42 flight area are ranging from 2 to 4. The aspect ratio is commonly between 210 2 to 6, with some observed extreme values higher than 10 (LeMone, 1973; Brown, 1980; 211 Etling & Brown, 1993). One of the common MABL parameter extracted from SAR mea-212 surements is its depth, assuming a fixed aspect ratio (Sikora et al., 1997; Young et al., 213 2000). We advise caution when using this estimation method given the spread in observed 214 aspect ratios and the spatial variability of MABL depths. 215



Figure 1. (a) Backscatter signal from SAR (σ_0), superimposed with the infrared brightness temperature from GOES and wind vector from WindSat for February 13, 2020. Time series of (b) and (c) along-wind σ_0 fluctuations in green and ATR 42 vertical velocity fluctuations (w') at $z \sim 300$ m in orange, (d) and (e) cross-wind σ'_0 in blue and w' in brown. The aircraft and SAR transects are shown by the different colored lines in map (a). (f) Organized structure length-scale (L_{OS}) estimates from the ATR 42, from SAR with 1D and 2D autocorrelations in the pink diamond area.



Figure 2. (a) SAR image with an organization criterion: cell / roll transitions (f < 0, 7) in orange and rolls ($f \ge 0.7$) in green. The black arrows are the WindSat wind and the red lines are the roll directions. Examples of $\sigma 0$ and associated 2D autocorrelation for (b) a cell / roll transition regime and (c) a well-established rolls regime. (d) Flatness versus the difference between the two estimates of structure directions, (e) difference between the structure direction and the wind direction and (f) organized structure lengthscale (L_{OS}) according to the ratio max(σ_0) / min(σ_0) over each tiles.

The 2D autocorrelation analysis, extended to 58 subdomains in the entire SAR im-216 age (Fig. 2a), allows us to explore the spatial variability of MABL structures. The sep-217 aration between cells and transition zones using the flatness criterion shows that con-218 vective rolls (green tiles) are the dominant type of clear sky MABL organization, given 219 the prevailing large-scale conditions of strong wind speed (Etling & Brown, 1993). Also, 220 the transition zones (orange tiles) are mainly located downwind of cloud boundaries. They 221 can be interpreted as buffer zones between the sub-cloud organization driven by the cold 222 pools and the clear-sky rolls, only slightly disturbed by the cloud activity. 223

224 When the occurring structures are clearly identified as convective rolls, the difference between the two independent estimates of roll orientation, Ψ and α , becomes low 225 (Fig. 2d). The roll direction estimate is hence robust, unlike cell/roll transition cases for 226 which defining a direction is hardly relevant. Over the study area, the wind direction from 227 WindSat, exhibits only slight variations, between 50° and 75° , characteristic of the es-228 tablished trade-wind flow. The roll axis is mostly oriented along the mean MABL wind, 229 with a slight positive bias and a variability that can reach $\pm 10^{\circ}$ (Fig. 2e), in line with 230 previous studies (Alpers & Brümmer, 1994; Atkinson & Wu Zhang, 1996; Wang et al., 231 2019). Over the 39 tiled areas detected as rolls, 22 have a well-marked periodicity scale 232 (\mathcal{L}_{OS}) , while others do not have a noticeable periodicity but are still stretched structures 233 along the mean wind direction. In order to evaluate the roll size evolution as a function 234 of the dynamic conditions, L_{OS} as a function of the ratio between the maximum and the 235 minimum of σ_0 over a each tile is shown in Fig. 2f. We assume that a higher modula-236 tion of the roughness is associated to stronger roll imprints: which is confirmed by the 237 results of Fig. 2f. 238

The WindSat wind speed estimates have low variability, with values between 10.0 239 m s⁻¹ and 11.0 m s⁻¹ and an accuracy of 0.2 m s⁻¹ (Zhang et al., 2018). Even if the 240 wind speed is almost constant, the roll sizes are variable. Convective rolls are preferen-241 tially encountered when shear-induced turbulence dominates (Etling & Brown, 1993: Stopa 242 et al., 2022). As the shear is proportional to the cube of the surface velocity, the wind 243 intensity is a key parameter in the development of these coherent structures but it is not 244 the exclusive source of roll size variability. Based only on one case study, these results 245 should be considered with caution. Such analysis should be applied on all the SAR wide 246 swath images available during the EUREC4A field campaign to document the diversity 247 of large-scale conditions. 248

²⁴⁹ 4 Shallow Convection and Cold Pools

Figure 3b illustrates the results of the cold pool identification method on an iso-250 lated feature in the south of the SAR swath. From the detected contours of the cold pool 251 (orange lines), an ellipse can be fitted (green circle), and morphological characteristics 252 of the detected feature can be derived, such as the size, the center of gravity, the flat-253 ness and the orientation of the ellipse. Also, dynamic properties can be inferred such as 254 the intensity of the gust front gradient (Fig. 3c), equal to $4.09 \times 10^{-2} \text{ dB km}^{-1}$ which 255 can be indicative of an active cold pool or the maximum of roughness (yellow contour 256 in Fig. 3a), equal to 4.16×10^{-1} dB. In SAR observations, the origin of such strong lo-257 calized σ_0 increase can be ambiguous between a rain signature (Alpers et al., 2016) or 258 a disturbance of the signal due to the hydrometeors in the upper part of a deep convec-259 tive system (Alpers et al., 2021). Only shallow convective clouds were encountered dur-260 ing EUREC4A, with cloud top around 2 km - 3 km and no ice hydrometeors are present. 261 Therefore, such roughness peaks is related to a splash zone, associated with rain cells. 262

The superposition of GOES T_B on SAR σ_0 at the same time (Fig. 3d) highlights the strong correlation between the cold pool imprint on the sea surface roughness and the cloud organization. Cloud and MABL dynamic processes are tightly coupled, with shallow convective clouds that remain mainly connected to the MABL. Based on the anal-



Figure 3. Identification method applied on an isolated cold pool and associated metrics. (a) Backscatter signal centered on the studied feature. (b) Results of the detection method (pink contours) and elliptical fit (green contours). (c) Transect across the cold pool along the red line shown in (a). (d) Brightness temperature superimposed on the SAR image at the same instant. (e) Backward tracking of the cloud cold core ($T_B \leq 285$ K).

ysis of the synoptic conditions, the large patch of T_B colder than 280 K in the northwest 267 of the cold pool, is due to high clouds with a north eastward advection typical of the at-268 mospheric layer above the south westward trade wind flow. There is no apparent signa-269 ture on the surface, these high clouds are completely decoupled of the MABL. However, 270 a relative minimum in T_B that can be associated to an ascending core can be observed 271 above the cold pool. In addition to the collocated SAR – GOES superposition in time, 272 a tracking of the cloud cold core $(T_B \leq 285 \text{ K})$ is performed, as shown in Fig. 3e. At 273 the SAR timing, the high altitude cloud layer interferes with the shallow convection clouds 274 but the distinction is noticeable as soon as we go back 5 min before. The progressive growth 275 of the cloud from its origin is observed. The 285 K level for this cloud was first detected 276 at 0810 UTC at about 61 km with a growth factor of $1.4 \text{ km}^2 \text{ min}^{-1}$ during the first hour 277 following its detection. This backward tracking allows to explore the origin of the cloud 278 entity and provides an estimate of the cold pool lifetime up to the present time, which 279 in this case is about 1 hour. 280

The application of the identification method on the entire large swath leads to the 281 detection of 68 cold pools (Fig. 4a). Most of them are located below the mesoscale shal-282 low clouds in the center of the image. Due to interactions and collisions between them, 283 these cold pools present a wide variability of size, shape and orientation. Some cold pools, 284 associated with isolated clouds, have a smaller variability of geometric parameters and 285 are generally smaller than the aggregated cold pools. The deduced morphological and 286 dynamic cold pool metrics allow us to explore the physical processes involved. In par-287 ticular, a correlation between the area of the cold pool and the maximum of σ_0 over the 288 cold pool is found (Fig. 4b). The more active and more precipitating clouds continue 289 to energetically supply the cold pool through rain evaporation, which leads to a spread-290 ing of the latter. Furthermore, these convective systems induce a high σ_0 maximum ei-291 ther related to splash zones directly due to the rain or to an increase of the wind inten-292 sity thanks to secondary circulations. The more active cold pools have a stronger rough-293 ness contrast (maximum of σ_0) with their environment and thus a sharper edge gradi-294 ent related to the gust front gradient intensity (Fig. 4e). 295

Cloud properties can be inferred from T_B and put in relation with cold pool prop-296 erties from SAR measurements. The parameter $\Delta_{\text{life}}T_B$ is the difference between the min-297 imum of T_B at the present time and the one since the first detection of the 285 K level, 298 over the followed cloud object. If $\Delta_{\text{life}}T_B$ is equal to 0, the T_B minimum is reached at 299 the SAR timing, and the higher $\Delta_{\text{life}}T_B$, the older the cloud peak intensity. A tendency 300 emerges, with less intense cold pool gust front gradients for higher $\Delta_{\text{life}}T_B$ (Fig. 4c). This 301 reflects a coupled loss of convective intensity both in the cloud layer and at the surface, 302 with potentially dissipating cold pools. The parameter $\Delta_{\text{shield}}T_B$ is defined, at each in-303 stant, as the difference between the averaged T_B of the cloud and the minimum of T_B 304 for this cloud. It provides insight into how active the cloud is, a small $\Delta_{\text{shield}}T_B$ indicates 305 the existence of cold cores and localized vertical developments. Higher σ_0 maxima are 306 associated to smaller $\Delta_{\text{shield}}T_B$ (Fig. 4d), which allow us to connect the more intense cold 307 pool imprint seen from below, at the surface to the intensity of the convective activity 308 seen from above. 309

310 5 Conclusions

Based on an EUREC4A case study, an extensive analysis of a wide swath SAR image has been performed with the combined use of in situ airborne measurement and geostationary brightness temperature. The SAR backscattered signal provides a quasi-instantaneous picture of the sea surface roughness. The signatures of clear-sky organization into convective rolls and cold pools below clouds have been analyzed.

The intercomparison of the observed structures on the sea surface roughness with 316 the MABL turbulent organization through the vertical velocity fluctuations measured 317 by the ATR 42 aircraft has allowed to consolidate the hypothesis of a significant rela-318 tionship between the surface small scales and the MABL processes. The SAR wide swath 319 has been divided into sub-domains of 25 x 25 km² to perform, on clear-sky areas, 2-D 320 autocorrelation of the high resolution roughness, enabling us to capture the spatial vari-321 ability MABL organization. Considering the observed high wind speed conditions, the 322 convective rolls were the predominant coherent structures. These rolls are almost ori-323 ented in the mean surface wind direction with characteristic scales between 2.0 and 3.9 324 km. The cell/roll transition zones have been mainly observed in the surrounding clear-325 sky environment of clouds. 326

Shallow convective systems have been investigated with an innovative approach, combining the SAR σ_0 with the geostationary cloud T_B in order to access the dynamics of clouds as seen from below and above. A segmentation method has been developed to detect the cold pools on σ_0 . The characteristics of the cold pools, based on geometrical or physically-based metrics are thus extracted. Within the SAR wide swath, a large



Figure 4. (a) Overview of the cold pool detection over the entire large swath, (b) σ_0 maximum versus the area, (c) $\Delta_{\text{life}}T_B$ versus the gust front gradient, (d) $\Delta_{\text{shield}}T_B$ versus the σ_0 maximum and (e) gust front gradient versus σ_0 maximum.

number of cold pools have been identified with significant variability in their properties.
The more active cold pools, associated with a higher MABL dynamic intensity, present
sharper edge roughness contrasts and higher roughness maxima. The intensity of the cloud
activity, with a pronounced vertical development, has been connected to the strong cold
pool imprint on sea surface roughness. Also, the relation between the decrease of convective intensity in the cloud layer and the dissipating cold pools at the surface has been
noted.

The new perspectives, presented here with a case study, on clear-sky turbulence 330 and shallow convection which can be extracted from a wide swath SAR image can pave 340 the way to improve our understanding of boundary layer processes from satellite obser-341 vations. The opportunity to jointly analyze characteristics of cold pools and related con-342 vection cloud organization, which can take a spectacular variety of forms and shapes (such 343 as isolated clouds, arcs, mesoscale systems, ...) should provide significant insights into 344 the mechanisms involved. The next step is to apply our approach to an ensemble of vary-345 ing conditions, targeting other regions and convective regimes, both shallow and deep 346 convection. The surrounding clear-sky environment and its interplay with the cloud sys-347 tems are also of great interest. 348

³⁴⁹ 6 Open Research

Sentinel-1 is part of the European space component of the Copernicus European 350 program. level-1 GRD data are free of charge and available on the Copernicus Open Ac-351 cess Hub (https://sentinels.copernicus.eu/web/sentinel/sentinel-data-access). 352 GOES-16 ABI Level 1b radiances are available at https://doi.org/10.7289/V5BV7DSR. 353 WindSat data are produced by Remote Sensing Systems and sponsored by the NASA 354 Earth Science MEaSURES DISCOVER Project and the NASA Earth Science Physical 355 Oceanography Program. RSS WindSat data are available at https://data.remss.com/ 356 windsat/. The EUREC4A turbulence dataset derived from the SAFIRE ATR 42 air-357 craft is available on the AERIS database (https://doi.org/10.25326/128). 358

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367 **References**

- 368Alpers, W., & Brümmer, B. (1994). Atmospheric boundary layer rolls observed by369the synthetic aperture radar aboard the ers-1 satellite. Journal of Geophys-370ical Research: Oceans, 99(C6), 12613-12621. doi: https://doi.org/10.1029/37194JC00421
- 372Alpers, W., Zhang, B., Mouche, A., Zeng, K., & Chan, P. W. (2016). Rain foot-373prints on c-band synthetic aperture radar images of the ocean revisited. Re-374mote Sensing of Environment, 187, 169-185. doi: https://doi.org/10.1016/j.rse375.2016.10.015
- Alpers, W., Zhao, Y., Mouche, A. A., & Chan, P. W. (2021). A note on radar signatures of hydrometeors in the melting layer as inferred from sentinel-1 sar data acquired over the ocean. *Remote Sensing of Environment*, 253, 112177. doi: https://doi.org/10.1016/j.rse.2020.112177

380	Atkinson, B. W., & Wu Zhang, J. (1996). Mesoscale shallow convection in the atmo-
381	sphere. Reviews of Geophysics, 34(4), 403-431. doi: https://doi.org/10.1029/
382	96RG02623
383	Atlas, D. (1994). Footprints of storms on the sea: A view from spaceborne synthetic $l = l + l + l + l + l + l + l + l + l + $
384	aperture radar. Journal of Geophysical Research: Oceans, 99(C4), 7961-7969.
385	doi: $nttps://doi.org/10.1029/94JC00250$
386	Ayet, A., Rascle, N., Chapron, B., Couvreux, F., & Terray, L. (2021). Un-
387	covering air-sea interaction in oceanic submessoscale frontal regions using high regolution stability observations US Clinan containing $10(1)$ doi:
388	https://doi.org/10.5065/when 0:03
389	Pony S. Lethon M. Delence I. Coutrie D. Etienne, I. C. Acmicegree F.
390	bony, S., Lothon, M., Deranoe, J., Courtis, F., Ettenne, JC., Aennisegger, F., Voral B (2022) $\operatorname{FUREC}^4\Lambda$ observations from the SAFIRE ATR 42
391	aircraft Earth System Science Data Discussions 2022 1-61 doi:
392	10 5194/essd-2021-459
304	Brient F Couvreux F Villefrancue N Bio C & Honnert B (2019 March)
395	Object-Oriented Identification of Coherent Structures in Large Eddy Simu-
396	lations: Importance of Downdrafts in Stratocumulus. <i>Geophysical Research</i>
397	Letters, 46(5), 2854–2864. doi: https://doi.org/10.1029/2018GL081499
398	Brilouet, PE., Durand, P., Canut, G., & Fourrié, N. (2020). Organized turbu-
399	lence in a cold-air outbreak: Evaluating a large-eddy simulation with respect
400	to airborne measurements. Boundary-Layer Meteorology, 175(1), 57–91. doi:
401	https://doi.org/10.1007/s10546-019-00499-4
402	Brilouet, PE., Lothon, M., Etienne, JC., Richard, P., Bony, S., Lernoult, J.,
403	Charoy, T. (2021) . The EUREC ⁴ A turbulence dataset derived from the
404	SAFIRE ATR 42 aircraft. Earth System Science Data Discussions, 2021,
405	1-28. doi: $10.5194/essd-2021-52$
406	Brown, R. A. (1980). Longitudinal instabilities and secondary flows in the plane-
407	tary boundary layer: A review. Reviews of Geophysics, 18(3), 683-697. doi:
408	https://doi.org/10.1029/RG018i003p00683
409	Chou, SH., & Ferguson, M. P. (1991). Heat fluxes and roll circulations over the
410	western gulf stream during an intense cold-air outbreak. Boundary-layer mete-
411	orology, 55(3), 255-281. doi: https://doi.org/10.1007/BF00122580
412	de Szoeke, S. P., Skyllingstad, E. D., Zuidema, P., & Chandra, A. S. (2017). Cold
413	pools and their influence on the tropical marine boundary layer. Journal of the
414	Atmospheric Sciences, 74(4), 1149 - 1168. doi: https://doi.org/10.1175/JAS-D
415	-16-0264.1
416	Etling, D., & Brown, R. A. (1993). Roll vortices in the planetary boundary layer: A
417	review. Boundary-Layer Meteorology, 65(3), 215–248. doi: https://doi.org/10
418	.1007/BF00705527
419	reng, Z., nagos, S., Kowe, A. K., Burleyson, U. D., Martini, M. N., & de Szoeke,
420	5. P. (2015). Mechanisms of convective cloud organization by cold pools
421	of Advances in Modeling Earth Systems 7(2) 357-381 doi: https://doi.org/
422	10 1002/2014MS000384
423	Garg P. Neshitt S. W. Lang, T. I. Priftis, G. Chronis, T. Thaver, I. D. &
424	Hence D A (2020) Identifying and characterizing tropical oceanic mesoscale
426	cold pools using spaceborne scatterometer winds. Journal of Geonhusical Re-
427	search: Atmospheres, 125(5), e2019JD031812. doi: https://doi.org/10.1029/
428	2019JD031812
429	Granero-Belinchon, C., Roux, S., Garnier, N., Tandeo, P., Chapron, B., & Mouche.
430	A. (2022). Two-dimensional structure functions to characterize convective rolls
431	in the marine atmospheric boundary layer from sentinel-1 SAR images.
432	doi: https://hal.archives-ouvertes.fr/hal-03576400
433	Khairoutdinov, M., & Randall, D. (2006). High-resolution simulation of shallow-
434	to-deep convection transition over land. Journal of the Atmospheric Sciences,

435	63(12), 3421 - 3436. doi: https://doi.org/10.1175/JAS3810.1
436	Kilpatrick, T. J., & Xie, SP. (2015). Ascat observations of downdrafts from
437	mesoscale convective systems. $Geophysical Research Letters, 42(6), 1951$ -
438	1958. Retrieved from https://agupubs.onlinelibrary.wiley.com/doi/abs/
439	10.1002/2015GL063025 doi: https://doi.org/10.1002/2015GL063025
440	Kudryavtsev, V., Akimov, D., Johannessen, J., & Chapron, B. (2005). On radar
441	imaging of current features: 1. model and comparison with observations. Jour-
442	nal of Geophysical Research: Oceans, 110(C7). doi: https://doi.org/10.1029/
443	2004JC002505
444	La, T. V., Messager, C., Honnorat, M., Sahl, R., Khenchaf, A., Channelliere, C., &
445	Lattes, P. (2020). Use of sentinel-1 c-band sar images for convective system
446	surface wind pattern detection. Journal of Applied Meteorology and Climatol-
447	ogy, 59(8), 1321 - 1332. doi: https://doi.org/10.1175/JAMC-D-20-0008.1
448	LeMone, M. A. (1973). The structure and dynamics of horizontal roll vortices in
449	the planetary boundary layer. Journal of Atmospheric Sciences, $30(6)$, 1077 -
450	1091. doi: https://doi.org/10.1175/1520-0469(1973)030 $\langle 1077:TSADOH \rangle 2.0.CO;$
451	2
452	Lemone, M. A. (1976). Modulation of turbulence energy by longitudinal rolls in
453	an unstable planetary boundary layer. Journal of Atmospheric Sciences, $33(7)$,
454	1308 - 1320. doi: https://doi.org/10.1175/1520-0469(1976)033 (1308:MOTEBL)
455	2.0.CO;2
456	Lima, M. A., & Wilson, J. W. (2008). Convective storm initiation in a moist tropi-
457	cal environment. Monthly Weather Review, 136(6), 1847 - 1864. doi: https://
458	doi.org/10.1175/2007MWR2279.1
459	Lohou, F., Druilhet, A., Campistron, B., Redelspergers, JL., & Saïd, F. (2000).
460	Numerical study of the impact of coherent structures on vertical transfers in
461	the atmospheric boundary layer. Boundary-layer meteorology, 97(3), 361–383.
462	doi: 10.1023/A:1002641728075
463	Mapes, B., Milliff, R., & Morzel, J. (2009). Composite life cycle of maritime trop-
464	ical mesoscale convective systems in scatterometer and microwave satellite
465	observations. Journal of the Atmospheric Sciences, $66(1)$, 199 - 208. doi:
466	https://doi.org/10.1175/2008JAS2746.1
467	Schmit, T. J., Gunshor, M. M., Menzel, W. P., Gurka, J. J., Li, J., & Bachmeier,
468	A. S. (2005). Introducing the next-generation advanced baseline imager on
469	goes-r. Bulletin of the American Meteorological Society, 86(8), 1079 - 1096.
470	doi: https://doi.org/10.1175/BAMS-86-8-1079
471	Sikora, T. D., Young, G. S., Shirer, H. N., & Chapman, R. D. (1997). Estimating
472	convective atmospheric boundary layer depth from microwave radar imagery
473	of the sea surface. Journal of Applied Meteorology, 36(7), 833 - 845. doi:
474	nttps://doi.org/10.1175/1520-0450(1997)036(0833:ECABLD)2.0.CO;2
475	Stevens, B., Bony, S., Farrell, D., Ament, F., Blyth, A., Fairall, C., Zöger, M.
476	(2021). EUREC ⁴ A. Earth System Science Data Discussions, 2021, 1–78. doi:
477	10.5194/essd-2021-18
478	Stopa, J. E., Wang, C., Vandemark, D., Foster, R., Mouche, A., & Chapron, B.
479	(2022). Automated global classification of surface layer stratification using
480	high-resolution sea surface roughness measurements by satellite synthetic
481	aperture radar. Geophysical Research Letters, $49(12)$, e2022GL098686. doi: https://doi.org/10.1020/2022GL008686
482	$\frac{1}{10000000000000000000000000000000000$
483	10rri, G., & Kuang, Z. (2019). Un cold pool collisions in tropical boundary lay-
484	ers. Geophysical Research Letters, 40(1), 399-407. doi: https://doi.org/10 1020/2018CI 080501
485	1025/20100L000001 Torri C. Kuong 7. Iz Tion V. (2015). Machanisma for convection triggering by
486	cold pools <i>Combusical Research Letters</i> 10(6) 1042 1050 doi: https://doi
487	org/10/1002/2015CL/0632227
488	.018/10.1002/201001003221

489	Vandemark, D., Mourad, P. D., Bailey, S. A., Crawford, T. L., Vogel, C. A., Sun, J.,
490	& Chapron, B. (2001). Measured changes in ocean surface roughness due to
491	atmospheric boundary layer rolls. Journal of Geophysical Research: Oceans,
492	106(C3), 4639-4654. doi: https://doi.org/10.1029/1999JC000051
493	Villefranque, N., Williamson, D., Couvreux, F., Hourdin, F., Gautrais, J., Fournier,
494	R., Volodina, V. (2020). Process-based climate model development har-
495	nessing machine learning: Iii. the representation of cumulus geometry and
496	their 3d radiative effects. Earth and Space Science Open Archive, 30. doi:
497	https://doi.org/10.1002/essoar.10505088.1
498	Vogel, R., Konow, H., Schulz, H., & Zuidema, P. (2021). A climatology of trade-
499	wind cumulus cold pools and their link to mesoscale cloud organization. Atmo-
500	spheric Chemistry and Physics, 21(21), 16609–16630. doi: https://doi.org/10
501	.5194/acp-21-16609-2021
502	Wang, C., Mouche, A., Foster, R. C., Vandemark, D. C., Stopa, J. E., Tandeo, P.,
503	Chapron, B. (2019). Characteristics of Marine Atmospheric Boundary
504	Layer Roll Vortices from Sentinel-1 Sar Wave Mode. IGARSS 2019 - 2019
505	IEEE International Geoscience and Remote Sensing Symposium, 7908-7911.
506	doi: https://doi.org/10.1109/IGARSS.2019.8900287
507	Wang, C., Vandemark, D., Mouche, A., Chapron, B., Li, H., & Foster, R. C. (2020).
508	An assessment of marine atmospheric boundary layer roll detection using
509	sentinel-1 sar data. Remote Sensing of Environment, 250, 112031. doi:
510	https://doi.org/10.1016/j.rse.2020.112031
511	Weckwerth, T. M., Wilson, J. W., & Wakimoto, R. M. (1996). Thermody-
512	namic variability within the convective boundary layer due to horizon-
513	tal convective rolls. Monthly Weather Review, 124(5), 769 - 784. doi:
514	$https://doi.org/10.1175/1520-0493(1996)124\langle 0769: TVWTCB\rangle 2.0.CO; 2000000000000000000000000000000000000$
515	Young, G. S., Sikora, T. D., & Winstead, N. S. (2000). Inferring marine atmo-
516	spheric boundary layer properties from spectral characteristics of satellite-
517	borne sar imagery. Monthly Weather Review, $128(5)$, $1506 - 1520$. doi:
518	10.1175/1520-0493(2000)128(1506:IMABLP)2.0.CO;2
519	Zhang, L., Shi, H., Wang, Z., Yu, H., Yin, X., & Liao, Q. (2018). Comparison of
520	wind speeds from spaceborne microwave radiometers with in situ observa-
521	tions and ecmwf data over the global ocean. Remote Sensing, $10(3)$. doi:
522	https://doi.org/10.3390/rs10030425
523	Zuidema, P., Li, Z., Hill, R. J., Bariteau, L., Rilling, B., Fairall, C., Hare, J.
524	(2012). On trade wind cumulus cold pools. Journal of the Atmospheric Sci-

⁵²⁵ ences, 69(1), 258 - 280. doi: https://doi.org/10.1175/JAS-D-11-0143.1

Clear-Sky Turbulence and Shallow Convection: New Insights Combining SAR Images, Satellite Brightness Temperature and In Situ Measurements

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Key Points:

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15	•	Atmospheric coherent structures (cells, rolls and cold pools) are systematically de-
16		tected and analysed in a high-resolution SAR image
17	•	Properties of rolls from SAR measurements are comparable with the turbulence
18		organization deduced from airborne data

 A diversity of cold pool geometrical and dynamical features is related to cloud intensity provided by satellite brightness temperature

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21 Abstract

The imprint of marine atmospheric boundary layer (MABL) dynamical structures on sea 22 surface roughness, as seen from Sentinel-1 Synthetic Aperture Radar (SAR) acquisitions, 23 is investigated. We focus on February 13th, 2020, a case study of the EUREC4A (Elu-24 cidating the role of clouds-circulation coupling in climate) field campaign. For clear sky 25 conditions, convective rolls and cells imprints on sea surface roughness is confirmed through 26 the intercomparison with MABL turbulent organization deduced from airborne measure-27 ments. A discretization of the SAR wide swath into $25 \ge 25 \text{ km}^2$ tiles allows us to cap-28 ture the spatial variability of the turbulence organization varying from rolls to cells. We 29 then objectively detect cold pools within the SAR image and combine them with geo-30 stationary brightness temperature. The geometrical or physically-based metrics of cold 31 pools are correlated to cloud properties. This provides a promising methodology to an-32 alyze the dynamics of convective systems as seen from below and above. 33

³⁴ Plain Language Summary

We propose an innovative approach to investigate the marine atmospheric bound-35 ary layer (MABL) dynamical structures by combining spaceborne Synthetic Aperture 36 Radar (SAR) images, brightness temperature (T_B) from the Geostationary Operational 37 Environmental Satellite (GOES) and in situ turbulence airborne measurements from the 38 EUREC4A field campaign. Focusing on February 13th, 2020, two types of atmospheric 39 processes are investigated: clear sky organizations and cold pools. The signature of at-40 mospheric coherent structures on sea surface roughness, especially convective rolls, is val-41 idated with respect to the turbulence measurements of the ATR 42 aircraft. The cold 42 pools are detected within the SAR image using a segmentation method. Cold pool char-43 acteristics such the size and the gust front intensity can then be directly derived from 44 the SAR image. The GOES images provide cloud field properties with the temporal di-45 mension. Exploring backward cloud evolution with respect to the SAR image timing ap-46 pears meaningful to catch the life cycle of cold pools and convective clouds from which 47 they originate. The application of this approach could pave the way to access the dy-48 namics of convective systems as seen from below and above, allowing to go one step fur-49 ther in the quantitative use of SAR images to investigate boundary layer processes. 50

51 **1** Introduction

Marine atmospheric boundary layer (MABL) dynamics plays a crucial role in the 52 mesoscale organization of convection. Among the different kinds of coherent organiza-53 tions occurring inside the MABL, three-dimensional cells and quasi-two-dimensional con-54 vective rolls are frequent (Etling & Brown, 1993; Weckwerth et al., 1996; Atkinson & Wu Zhang, 55 1996; Wang et al., 2020). The roll wavelengths are in the range of spectral scales that 56 contribute the most to turbulence kinetic energy (Lemone, 1976). It leads to a substan-57 tial impact on the vertical transport of heat and moisture (Chou & Ferguson, 1991; Etling 58 & Brown, 1993; Brilouet et al., 2020), which is still not completely understood. Another 59 important kind of MABL coherent structures are the cold pools which are key compo-60 nents for the life cycle of convection and cloud organization. Cold pools are generated 61 by the downward motion of air mass cooled by rain drop evaporation. At the surface, 62 it spreads out horizontally to form a gust front with an enhanced wind intensity. Pre-63 vious studies have highlighted the relationships between cold pools, their accompany-64 ing moisture distribution and related cloud cover (Khairoutdinov & Randall, 2006; Zuidema 65 et al., 2012; Torri et al., 2015). Cold pools can, in their core, inhibit the development 66 of new convective cells due to their stabilizing effects. However, they can also dynam-67 ically trigger, at their edges, new convective cells, through wind convergence and lifting 68 or thermodynamically by accumulation of moisture (Lima & Wilson, 2008; Feng et al., 69

2015; Torri & Kuang, 2019). It is then fundamental to improve the monitoring of cold
 pools and their interplay with convection systems.

To improve understanding of MABL coherent structures, accurate measurements 72 are needed. Most observational studies are based on in-situ data, including shipboard 73 (e.g., Zuidema et al., 2012; de Szoeke et al., 2017) or shore-based measurements (e.g., 74 Vogel et al., 2021): those provide thermodynamical and dynamical measurements, with 75 however spatially-limited sampling. For shallow convection MABL coherent structures, 76 spatial sampling could be obtained with high-resolution (< 100 m) spaceborne obser-77 78 vations of sea-surface roughness, related to surface wind and stability variations. In particular, SAR backscatter is such a high resolution ($\sim 10 \text{ m} - 100 \text{ m}$) measurement, avail-79 able during day and night, regardless of weather conditions and cloud cover. Coherent 80 structures induce a contrasts of SAR backscatter, which are however difficult to extract 81 since they are multiscale and superimpose with other (oceanic) signatures (Kudryavtsev 82 et al., 2005; Avet et al., 2021). Atlas (1994) was a pioneer in highlighting the potential 83 signature of a storm microburst on the sea surface. Further work focused on convective 84 rolls (Alpers & Brümmer, 1994; Young et al., 2000; Vandemark et al., 2001; Wang et al., 85 2020), their link with stratification Stopa et al. (2022) and on deep convection cold pools 86 from SAR winds (La et al., 2020) or from scatterometer winds (Mapes et al., 2009; Kil-87 patrick & Xie, 2015; Garg et al., 2020). 88

Systematic and quantitative extraction of high resolution information on MABL 89 properties in SAR measurements is still a challenge, due to overlapping geophysical pro-90 cesses which require complementary information. The aim of this paper is thus to present 91 an innovative approach based on a combined use of SAR and geostationary satellite data 92 to study the interplay between the MABL and convective organization. By doing so we 93 (i) show how SAR clear-sky convection compares to in situ measurements and (ii) anal-94 yse the morphological and dynamical properties of cold pools, related to moist convec-95 tion. 96

The paper is organized as follows: the data and methods are presented in Section
Section 3 is devoted to the clear-sky turbulence and Section 4 concerns the cold pool detection and characterization. The last section concludes the paper with a discussion of the main results.

¹⁰¹ 2 Data and Methods

2.1 Data

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We used C-band SAR images from the Sentinel-1 satellite, which probes the sea surface through clouds, in interferometric wide swath (IW) mode. Each image is 400km wide with effective resolution of 100 m, allowing multiscale sampling of cold pools. The C-band backscatter signal (σ_0) is sensitive to the slope distribution of centimetrescale wind-wave, with short adjustment timescales to changes in surface wind (Kudryavtsev et al., 2005). Backscatter contrasts can thus be related to surface signatures of MABL processes (Ayet et al., 2021).

To obtain information on the cloud layer, we used infrared brightness temperature (T_B) provided by the Advanced Baseline Imager (ABI, Schmit et al., 2005) onboard GOES-16 (Geostationary Operational Environmental Satellite). In the region of interest, the Atlantic trades, cloud cover mainly consists of low-level clouds which facilitates the interpretation of the satellite images can be detected with the brightness temperature despite their small temperature differences with the sea surface. The spatial resolution is 2 km and temporal resolution 10 min. In addition, we used wind speed and direction estimated from WindSat, a polarimetric microwave radiometer operated on the Coriolis satellite, with a spatial resolution of 25 km.

Finally, we used in situ observations from the EUREC4A campaign (Elucidating the role of clouds-circulation coupling in climate), which took place over the Western part of the Tropical Atlantic near Barbados, in Jan-Feb 2020 (Stevens et al., 2021). In particular, the SAFIRE ATR 42 aircraft samples the atmosphere around the cloud-base level and at different heights in the subcloud layer (Bony et al., 2022). Among the collected data, we will focus on the vertical velocity turbulent fluctuations (w') deduced from a five-hole radome nose with a spatial resolution of around 4 m (Brilouet et al., 2021).

¹²⁷ Here, we focused on February 13th 2020, with the best spatial and temporal over-¹²⁸ lapping between ATR 42 and Sentinel-1 acquisition: an ATR 42 flight between 0735 UTC ¹²⁹ and 1152 UTC (take-off and landing times) and a Sentinel-1 crossing around 0935 UTC, ¹³⁰ about 100 km away from the aircraft track (see Fig. 1a). Superimposing T_B and σ_0 shows ¹³¹ distinct atmospheric processes: clear-sky convective rolls and cold pools below clouds.

132 **2.2 Methods**

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2.2.1 Characteristics from two-dimensional autocorrelation

Because turbulent processes are highly variable in time and space, as well as very 134 sensitive to their dynamical environment, the wide swath SAR image has been split into 135 $25 \times 25 \text{ km}^2$ tiles (Fig. 2a). This is a good compromise with sub-domains large enough 136 to properly sample coherent structures and small enough to be homogeneous. This is con-137 sistent with previous studies based on 20×20 km² Wave Mode SAR acquisitions which 138 focused on convective roll analysis (Wang et al., 2019; Granero-Belinchon et al., 2022). 139 Using the GOES image, tiles with cloudy pixels were removed from the clear-sky tur-140 bulence analysis (blue tiles in Fig. 2a). Based on in situ EUREC4A data, a T_B thresh-141 old of 292 K has been chosen : colder pixels are considered as cloudy. 142

Two-dimensional autocorrelation is calculated over each 25×25 km² tiled area. 143 Following the methodology detailed in Granero-Belinchon et al. (2022), we first estimate 144 the longitudinal axis of the coherent structures (noted hereafter as Ψ , Fig. 2b,c solid red 145 line). Then, following Lohou et al. (2000), the integral lengthscale L_E is estimated for 146 each angle θ and an elliptical fit is applied to the resulting polar curve (red ellipse in Figs. 2b,c). 147 The type of organization is then diagnosed from the flatness parameter f of the ellipse. 148 A threshold is empirically chosen to distinguish rolls ($f \ge 0.7$, green tiles in Fig. 2a 149 and Fig. 2c) from the transition between rolls and cells (f < 0.7, orange tiles in Fig. 2a 150 and Fig. 2b). When L_E anisotropy is strong enough, the orientation of the major axis 151 provides a second estimate of the roll direction (noted hereafter as α , dashed purple line 152 in Figs. 2b,c). Moreover, for roll tiles, the organized structure lengthscale L_{OS} is defined 153 as the autocorrelation secondary local maximum along the direction perpendicular to 154 the structures (blue line in Fig.2c): it corresponds to the mean transverse lengthscale 155 of the rolls. 156

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2.2.2 Identification and characterization of cold pools

To analyse the morphological and dynamical characteristics of cold pool signatures 158 on SAR, we developed an objective identification method, validated over the case study. 159 An example a cold pool signature is given in Fig. 3a. The method makes the physical 160 161 assumption that the pattern of a cold pool consists of strong positive gradients at its forward and backward edges (blue zones in Fig. 3c) and a rather smooth central area with 162 a negative gradient (green zone in Fig. 3c). This corresponds to the gust front and to 163 the horizontal wind divergence associated to the downdraft, respectively. The edges are 164 first detected by computing increments on the gaussian-filtered backscatter signal and 165

selecting pixels with increments larger than twice the increment standard deviation. Individual cold pools objects (orange contours in Fig. 3b) are then defined by grouping edges
together if a zone of negative increment exists between them (e.g., Fig. 3c, the blue edges
are part of the same cold pool because of the green zone with negative increment). For
the SAR image studied here, this objective methodology captures most of the cold pools,
as validated visually (Fig. 4a). Future work will focus on extending this method systematically to other SAR images.

Detected cold pools were contextualized with an object identification of cloud structures within GOES T_B . The method is described in Brient et al. (2019) and Villefranque et al. (2020): it uses the watershed algorithm in space and time (3D) with two thresholds on the cloud object ($T_B \leq 292$ K) and on the cold cores ($T_B \leq 285$ K), allowing to follow the identified object in time (contours in Fig. 3e). This method thus connects the instantaneous snapshot of cold pools with the temporal evolution of the cloud organization and individual cloud life cycle.

¹⁸⁰ 3 Clear-Sky Turbulence

The ability of SAR to retrieve MABL turbulence activity is assessed using airborne 181 observations. In spite of a non perfect overlap between airborne and spaceborne obser-182 vations, two transects were made in the SAR image to mimic the aircraft legs (Fig. 1a). 183 We assume the stationarity of the turbulent field and an advection with the mean wind 184 direction of the coherent structures contained in the SAR image over the ATR 42 area. 185 The direction of the two SAR transect blue and green lines) was chosen to have the same 186 orientation with respect to WindSat wind than the ATR42 (brown and orange line). The 187 time series of w' measured by the ATR 42 and σ_0 fluctuations highlight similar signal 188 dynamics (Fig. 1b). For the two sampling directions, signals differ significantly: the pe-189 riodicity of the sampled eddies is significantly smaller for the transversal legs (blue and 190 brown) than longitudinal ones (green and orange). This emphasizes the nonaxisymmet-191 ric behavior of both turbulence and roughness fields. 192

To document the uncertainties associated to the unidirectional nature of the in situ 193 airborne sampling, a set of 40 1D-SAR transects is extracted (contained within the pink 194 parallelepipede in Fig. 1a). The L_{OS} computed on this set show a substantial variabil-195 ity between 1.10 km and 4.71 km with an average value of 2.65 km (see the boxplot in 196 Fig. 1f). The estimates made on the ATR legs at 300 m and 600 m are 1.60 km and 2.00 m197 km, respectively, are well within this range of values covered by the 1D-SAR estimates. 198 This result is in line with the study of Vandemark et al. (2001) and consolidates the hy-199 pothesis of a sea surface roughness imprint from MABL coherent structures. One of the 200 SAR strength is to provide a 2D-field of roughness which allows for 2D autocorrelation 201 analysis. In the present case, the estimate of L_{OS} over the pink area from the 2D au-202 tocorrelation, 2.90 km, is well consistent with the 1D SAR range of values and with in 203 situ estimates. In the SAR subdomain considered here, a dropsonde launched from the 204 HALO aircraft close to the SAR passage timing reveals a well-mixed MABL with a depth 205 of $z_i = 570$ m. It leads to an aspect ratio of 5.1. For ATR42 measurements, the MABL 206 depth estimates are variable, ranging from $z_i = 500$ m to $z_i = 800$ m. This variabil-207 ity can be driven by dry air intrusions from the troposphere or by secondary circulation 208 dynamics generated by the mesoscale cloud system. Therefore, the aspect ratio estimates 209 in the ATR42 flight area are ranging from 2 to 4. The aspect ratio is commonly between 210 2 to 6, with some observed extreme values higher than 10 (LeMone, 1973; Brown, 1980; 211 Etling & Brown, 1993). One of the common MABL parameter extracted from SAR mea-212 surements is its depth, assuming a fixed aspect ratio (Sikora et al., 1997; Young et al., 213 2000). We advise caution when using this estimation method given the spread in observed 214 aspect ratios and the spatial variability of MABL depths. 215



Figure 1. (a) Backscatter signal from SAR (σ_0), superimposed with the infrared brightness temperature from GOES and wind vector from WindSat for February 13, 2020. Time series of (b) and (c) along-wind σ_0 fluctuations in green and ATR 42 vertical velocity fluctuations (w') at $z \sim 300$ m in orange, (d) and (e) cross-wind σ'_0 in blue and w' in brown. The aircraft and SAR transects are shown by the different colored lines in map (a). (f) Organized structure length-scale (L_{OS}) estimates from the ATR 42, from SAR with 1D and 2D autocorrelations in the pink diamond area.



Figure 2. (a) SAR image with an organization criterion: cell / roll transitions (f < 0, 7) in orange and rolls ($f \ge 0.7$) in green. The black arrows are the WindSat wind and the red lines are the roll directions. Examples of $\sigma 0$ and associated 2D autocorrelation for (b) a cell / roll transition regime and (c) a well-established rolls regime. (d) Flatness versus the difference between the two estimates of structure directions, (e) difference between the structure direction and the wind direction and (f) organized structure lengthscale (L_{OS}) according to the ratio max(σ_0) / min(σ_0) over each tiles.

The 2D autocorrelation analysis, extended to 58 subdomains in the entire SAR im-216 age (Fig. 2a), allows us to explore the spatial variability of MABL structures. The sep-217 aration between cells and transition zones using the flatness criterion shows that con-218 vective rolls (green tiles) are the dominant type of clear sky MABL organization, given 219 the prevailing large-scale conditions of strong wind speed (Etling & Brown, 1993). Also, 220 the transition zones (orange tiles) are mainly located downwind of cloud boundaries. They 221 can be interpreted as buffer zones between the sub-cloud organization driven by the cold 222 pools and the clear-sky rolls, only slightly disturbed by the cloud activity. 223

224 When the occurring structures are clearly identified as convective rolls, the difference between the two independent estimates of roll orientation, Ψ and α , becomes low 225 (Fig. 2d). The roll direction estimate is hence robust, unlike cell/roll transition cases for 226 which defining a direction is hardly relevant. Over the study area, the wind direction from 227 WindSat, exhibits only slight variations, between 50° and 75° , characteristic of the es-228 tablished trade-wind flow. The roll axis is mostly oriented along the mean MABL wind, 229 with a slight positive bias and a variability that can reach $\pm 10^{\circ}$ (Fig. 2e), in line with 230 previous studies (Alpers & Brümmer, 1994; Atkinson & Wu Zhang, 1996; Wang et al., 231 2019). Over the 39 tiled areas detected as rolls, 22 have a well-marked periodicity scale 232 (\mathcal{L}_{OS}) , while others do not have a noticeable periodicity but are still stretched structures 233 along the mean wind direction. In order to evaluate the roll size evolution as a function 234 of the dynamic conditions, L_{OS} as a function of the ratio between the maximum and the 235 minimum of σ_0 over a each tile is shown in Fig. 2f. We assume that a higher modula-236 tion of the roughness is associated to stronger roll imprints: which is confirmed by the 237 results of Fig. 2f. 238

The WindSat wind speed estimates have low variability, with values between 10.0 239 m s⁻¹ and 11.0 m s⁻¹ and an accuracy of 0.2 m s⁻¹ (Zhang et al., 2018). Even if the 240 wind speed is almost constant, the roll sizes are variable. Convective rolls are preferen-241 tially encountered when shear-induced turbulence dominates (Etling & Brown, 1993: Stopa 242 et al., 2022). As the shear is proportional to the cube of the surface velocity, the wind 243 intensity is a key parameter in the development of these coherent structures but it is not 244 the exclusive source of roll size variability. Based only on one case study, these results 245 should be considered with caution. Such analysis should be applied on all the SAR wide 246 swath images available during the EUREC4A field campaign to document the diversity 247 of large-scale conditions. 248

²⁴⁹ 4 Shallow Convection and Cold Pools

Figure 3b illustrates the results of the cold pool identification method on an iso-250 lated feature in the south of the SAR swath. From the detected contours of the cold pool 251 (orange lines), an ellipse can be fitted (green circle), and morphological characteristics 252 of the detected feature can be derived, such as the size, the center of gravity, the flat-253 ness and the orientation of the ellipse. Also, dynamic properties can be inferred such as 254 the intensity of the gust front gradient (Fig. 3c), equal to $4.09 \times 10^{-2} \text{ dB km}^{-1}$ which 255 can be indicative of an active cold pool or the maximum of roughness (yellow contour 256 in Fig. 3a), equal to 4.16×10^{-1} dB. In SAR observations, the origin of such strong lo-257 calized σ_0 increase can be ambiguous between a rain signature (Alpers et al., 2016) or 258 a disturbance of the signal due to the hydrometeors in the upper part of a deep convec-259 tive system (Alpers et al., 2021). Only shallow convective clouds were encountered dur-260 ing EUREC4A, with cloud top around 2 km - 3 km and no ice hydrometeors are present. 261 Therefore, such roughness peaks is related to a splash zone, associated with rain cells. 262

The superposition of GOES T_B on SAR σ_0 at the same time (Fig. 3d) highlights the strong correlation between the cold pool imprint on the sea surface roughness and the cloud organization. Cloud and MABL dynamic processes are tightly coupled, with shallow convective clouds that remain mainly connected to the MABL. Based on the anal-



Figure 3. Identification method applied on an isolated cold pool and associated metrics. (a) Backscatter signal centered on the studied feature. (b) Results of the detection method (pink contours) and elliptical fit (green contours). (c) Transect across the cold pool along the red line shown in (a). (d) Brightness temperature superimposed on the SAR image at the same instant. (e) Backward tracking of the cloud cold core ($T_B \leq 285$ K).

ysis of the synoptic conditions, the large patch of T_B colder than 280 K in the northwest 267 of the cold pool, is due to high clouds with a north eastward advection typical of the at-268 mospheric layer above the south westward trade wind flow. There is no apparent signa-269 ture on the surface, these high clouds are completely decoupled of the MABL. However, 270 a relative minimum in T_B that can be associated to an ascending core can be observed 271 above the cold pool. In addition to the collocated SAR – GOES superposition in time, 272 a tracking of the cloud cold core $(T_B \leq 285 \text{ K})$ is performed, as shown in Fig. 3e. At 273 the SAR timing, the high altitude cloud layer interferes with the shallow convection clouds 274 but the distinction is noticeable as soon as we go back 5 min before. The progressive growth 275 of the cloud from its origin is observed. The 285 K level for this cloud was first detected 276 at 0810 UTC at about 61 km with a growth factor of $1.4 \text{ km}^2 \text{ min}^{-1}$ during the first hour 277 following its detection. This backward tracking allows to explore the origin of the cloud 278 entity and provides an estimate of the cold pool lifetime up to the present time, which 279 in this case is about 1 hour. 280

The application of the identification method on the entire large swath leads to the 281 detection of 68 cold pools (Fig. 4a). Most of them are located below the mesoscale shal-282 low clouds in the center of the image. Due to interactions and collisions between them, 283 these cold pools present a wide variability of size, shape and orientation. Some cold pools, 284 associated with isolated clouds, have a smaller variability of geometric parameters and 285 are generally smaller than the aggregated cold pools. The deduced morphological and 286 dynamic cold pool metrics allow us to explore the physical processes involved. In par-287 ticular, a correlation between the area of the cold pool and the maximum of σ_0 over the 288 cold pool is found (Fig. 4b). The more active and more precipitating clouds continue 289 to energetically supply the cold pool through rain evaporation, which leads to a spread-290 ing of the latter. Furthermore, these convective systems induce a high σ_0 maximum ei-291 ther related to splash zones directly due to the rain or to an increase of the wind inten-292 sity thanks to secondary circulations. The more active cold pools have a stronger rough-293 ness contrast (maximum of σ_0) with their environment and thus a sharper edge gradi-294 ent related to the gust front gradient intensity (Fig. 4e). 295

Cloud properties can be inferred from T_B and put in relation with cold pool prop-296 erties from SAR measurements. The parameter $\Delta_{\text{life}}T_B$ is the difference between the min-297 imum of T_B at the present time and the one since the first detection of the 285 K level, 298 over the followed cloud object. If $\Delta_{\text{life}}T_B$ is equal to 0, the T_B minimum is reached at 299 the SAR timing, and the higher $\Delta_{\text{life}}T_B$, the older the cloud peak intensity. A tendency 300 emerges, with less intense cold pool gust front gradients for higher $\Delta_{\text{life}}T_B$ (Fig. 4c). This 301 reflects a coupled loss of convective intensity both in the cloud layer and at the surface, 302 with potentially dissipating cold pools. The parameter $\Delta_{\text{shield}}T_B$ is defined, at each in-303 stant, as the difference between the averaged T_B of the cloud and the minimum of T_B 304 for this cloud. It provides insight into how active the cloud is, a small $\Delta_{\text{shield}}T_B$ indicates 305 the existence of cold cores and localized vertical developments. Higher σ_0 maxima are 306 associated to smaller $\Delta_{\text{shield}}T_B$ (Fig. 4d), which allow us to connect the more intense cold 307 pool imprint seen from below, at the surface to the intensity of the convective activity 308 seen from above. 309

310 5 Conclusions

Based on an EUREC4A case study, an extensive analysis of a wide swath SAR image has been performed with the combined use of in situ airborne measurement and geostationary brightness temperature. The SAR backscattered signal provides a quasi-instantaneous picture of the sea surface roughness. The signatures of clear-sky organization into convective rolls and cold pools below clouds have been analyzed.

The intercomparison of the observed structures on the sea surface roughness with 316 the MABL turbulent organization through the vertical velocity fluctuations measured 317 by the ATR 42 aircraft has allowed to consolidate the hypothesis of a significant rela-318 tionship between the surface small scales and the MABL processes. The SAR wide swath 319 has been divided into sub-domains of 25 x 25 km² to perform, on clear-sky areas, 2-D 320 autocorrelation of the high resolution roughness, enabling us to capture the spatial vari-321 ability MABL organization. Considering the observed high wind speed conditions, the 322 convective rolls were the predominant coherent structures. These rolls are almost ori-323 ented in the mean surface wind direction with characteristic scales between 2.0 and 3.9 324 km. The cell/roll transition zones have been mainly observed in the surrounding clear-325 sky environment of clouds. 326

Shallow convective systems have been investigated with an innovative approach, combining the SAR σ_0 with the geostationary cloud T_B in order to access the dynamics of clouds as seen from below and above. A segmentation method has been developed to detect the cold pools on σ_0 . The characteristics of the cold pools, based on geometrical or physically-based metrics are thus extracted. Within the SAR wide swath, a large



Figure 4. (a) Overview of the cold pool detection over the entire large swath, (b) σ_0 maximum versus the area, (c) $\Delta_{\text{life}}T_B$ versus the gust front gradient, (d) $\Delta_{\text{shield}}T_B$ versus the σ_0 maximum and (e) gust front gradient versus σ_0 maximum.

number of cold pools have been identified with significant variability in their properties.
The more active cold pools, associated with a higher MABL dynamic intensity, present
sharper edge roughness contrasts and higher roughness maxima. The intensity of the cloud
activity, with a pronounced vertical development, has been connected to the strong cold
pool imprint on sea surface roughness. Also, the relation between the decrease of convective intensity in the cloud layer and the dissipating cold pools at the surface has been
noted.

The new perspectives, presented here with a case study, on clear-sky turbulence 330 and shallow convection which can be extracted from a wide swath SAR image can pave 340 the way to improve our understanding of boundary layer processes from satellite obser-341 vations. The opportunity to jointly analyze characteristics of cold pools and related con-342 vection cloud organization, which can take a spectacular variety of forms and shapes (such 343 as isolated clouds, arcs, mesoscale systems, ...) should provide significant insights into 344 the mechanisms involved. The next step is to apply our approach to an ensemble of vary-345 ing conditions, targeting other regions and convective regimes, both shallow and deep 346 convection. The surrounding clear-sky environment and its interplay with the cloud sys-347 tems are also of great interest. 348

³⁴⁹ 6 Open Research

Sentinel-1 is part of the European space component of the Copernicus European 350 program. level-1 GRD data are free of charge and available on the Copernicus Open Ac-351 cess Hub (https://sentinels.copernicus.eu/web/sentinel/sentinel-data-access). 352 GOES-16 ABI Level 1b radiances are available at https://doi.org/10.7289/V5BV7DSR. 353 WindSat data are produced by Remote Sensing Systems and sponsored by the NASA 354 Earth Science MEaSURES DISCOVER Project and the NASA Earth Science Physical 355 Oceanography Program. RSS WindSat data are available at https://data.remss.com/ 356 windsat/. The EUREC4A turbulence dataset derived from the SAFIRE ATR 42 air-357 craft is available on the AERIS database (https://doi.org/10.25326/128). 358

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367 References

- 368Alpers, W., & Brümmer, B. (1994). Atmospheric boundary layer rolls observed by369the synthetic aperture radar aboard the ers-1 satellite. Journal of Geophys-370ical Research: Oceans, 99(C6), 12613-12621. doi: https://doi.org/10.1029/37194JC00421
- 372Alpers, W., Zhang, B., Mouche, A., Zeng, K., & Chan, P. W. (2016).Rain foot-373prints on c-band synthetic aperture radar images of the ocean revisited.Re-374mote Sensing of Environment, 187, 169-185.doi: https://doi.org/10.1016/j.rse375.2016.10.015
- Alpers, W., Zhao, Y., Mouche, A. A., & Chan, P. W. (2021). A note on radar signatures of hydrometeors in the melting layer as inferred from sentinel-1 sar data acquired over the ocean. *Remote Sensing of Environment*, 253, 112177. doi: https://doi.org/10.1016/j.rse.2020.112177

380	Atkinson, B. W., & Wu Zhang, J. (1996). Mesoscale shallow convection in the atmo-
381	sphere. Reviews of Geophysics, 34(4), 403-431. doi: https://doi.org/10.1029/
382	96 RG02623
383	Atlas, D. (1994). Footprints of storms on the sea: A view from spaceborne synthetic
384	aperture radar. Journal of Geophysical Research: Oceans, 99(C4), 7961-7969.
385	doi: https://doi.org/10.1029/94JC00250
386	Ayet, A., Rascle, N., Chapron, B., Couvreux, F., & Terray, L. (2021). Un-
387	covering air-sea interaction in oceanic submesoscale frontal regions using
388	high-resolution satellite observations. US Clivar variations, $19(1)$. doi:
389	Dany S. Lethan M. Delanoï I. Coutria D. Etianna, I.C. Astriagonan E.
390	Bony, S., Lotnon, M., Delanoe, J., Coutris, P., Etlenne, JC., Aemisegger, F., Voral P. (2022) FUREC ⁴ A observations from the SAFIPE ATP 42
391	aircraft Earth System Science Data Discussions 2022 1-61 doi:
392	10.5194/essd-2021-459
204	Brient F Couvreux F Villefrancue N Bio C & Honnert B (2019 March)
395	Object-Oriented Identification of Coherent Structures in Large Eddy Simu-
396	lations: Importance of Downdrafts in Stratocumulus. <i>Geophysical Research</i>
397	Letters, 46(5), 2854–2864. doi: https://doi.org/10.1029/2018GL081499
398	Brilouet, PE., Durand, P., Canut, G., & Fourrié, N. (2020). Organized turbu-
399	lence in a cold-air outbreak: Evaluating a large-eddy simulation with respect
400	to airborne measurements. Boundary-Layer Meteorology, $175(1)$, $57-91$. doi:
401	https://doi.org/10.1007/s10546-019-00499-4
402	Brilouet, PE., Lothon, M., Etienne, JC., Richard, P., Bony, S., Lernoult, J.,
403	Charoy, T. (2021) . The EUREC ⁴ A turbulence dataset derived from the
404	SAFIRE ATR 42 aircraft. Earth System Science Data Discussions, 2021,
405	1-28. doi: $10.5194/essd-2021-52$
406	Brown, R. A. (1980). Longitudinal instabilities and secondary flows in the plane-
407	tary boundary layer: A review. <i>Reviews of Geophysics</i> , 18(3), 683-697. doi:
408	$\frac{1}{1001} = \frac{1}{1001} + 1$
409	Chou, SH., & Ferguson, M. P. (1991). Heat fluxes and roll circulations over the
410	western gun stream during an intense cold-air outbreak. Downauty-layer mele-
411	de Szoeke S P Skyllingstad E D Zuidema P & Chandra A S (2017) Cold
412	pools and their influence on the tropical marine boundary layer <i>Journal of the</i>
414	Atmospheric Sciences, 74(4), 1149 - 1168, doi: https://doi.org/10.1175/JAS-D
415	-16-0264.1
416	Etling, D., & Brown, R. A. (1993). Roll vortices in the planetary boundary layer: A
417	review. Boundary-Layer Meteorology, 65(3), 215–248. doi: https://doi.org/10
418	.1007/BF00705527
419	Feng, Z., Hagos, S., Rowe, A. K., Burleyson, C. D., Martini, M. N., & de Szoeke,
420	S. P. (2015). Mechanisms of convective cloud organization by cold pools
421	over tropical warm ocean during the amie/dynamo field campaign. Journal
422	of Advances in Modeling Earth Systems, 7(2), 357-381. doi: https://doi.org/
423	10.1002/2014MS000384
424	Garg, P., Nesbitt, S. W., Lang, T. J., Priftis, G., Chronis, T., Thayer, J. D., &
425	nence, D. A. (2020). Identifying and characterizing tropical oceanic mesoscale
426	search. Atmospheres 195(5) 2010 ID031812 doi: https://doi.org/10.1020/
421	2019.JD031812
420	Granero-Belinchon C Boux S Garnier N Tandeo P Chapron B & Mouche
430	A. (2022). Two-dimensional structure functions to characterize convective rolls
431	in the marine atmospheric boundary layer from sentinel-1 SAR images.
432	doi: https://hal.archives-ouvertes.fr/hal-03576400
433	Khairoutdinov, M., & Randall, D. (2006). High-resolution simulation of shallow-
434	to-deep convection transition over land. Journal of the Atmospheric Sciences,

435	63(12), 3421 - 3436. doi: https://doi.org/10.1175/JAS3810.1
436	Kilpatrick, T. J., & Xie, SP. (2015). Ascat observations of downdrafts from
437	mesoscale convective systems. $Geophysical Research Letters, 42(6), 1951$ -
438	1958. Retrieved from https://agupubs.onlinelibrary.wiley.com/doi/abs/
439	10.1002/2015GL063025 doi: https://doi.org/10.1002/2015GL063025
440	Kudryavtsev, V., Akimov, D., Johannessen, J., & Chapron, B. (2005). On radar
441	imaging of current features: 1. model and comparison with observations. Jour-
442	nal of Geophysical Research: Oceans, 110(C7). doi: https://doi.org/10.1029/
443	2004JC002505
444	La, T. V., Messager, C., Honnorat, M., Sahl, R., Khenchaf, A., Channelliere, C., &
445	Lattes, P. (2020). Use of sentinel-1 c-band sar images for convective system
446	surface wind pattern detection. Journal of Applied Meteorology and Climatol-
447	ogy, 59(8), 1321 - 1332. doi: https://doi.org/10.1175/JAMC-D-20-0008.1
448	LeMone, M. A. (1973). The structure and dynamics of horizontal roll vortices in
449	the planetary boundary layer. Journal of Atmospheric Sciences, $30(6)$, 1077 -
450	1091. doi: https://doi.org/10.1175/1520-0469(1973)030 $\langle 1077:TSADOH \rangle 2.0.CO;$
451	2
452	Lemone, M. A. (1976). Modulation of turbulence energy by longitudinal rolls in
453	an unstable planetary boundary layer. Journal of Atmospheric Sciences, $33(7)$,
454	1308 - 1320. doi: https://doi.org/10.1175/1520-0469(1976)033 (1308:MOTEBL)
455	2.0.CO;2
456	Lima, M. A., & Wilson, J. W. (2008). Convective storm initiation in a moist tropi-
457	cal environment. Monthly Weather Review, 136(6), 1847 - 1864. doi: https://
458	doi.org/10.1175/2007MWR2279.1
459	Lohou, F., Druilhet, A., Campistron, B., Redelspergers, JL., & Saïd, F. (2000).
460	Numerical study of the impact of coherent structures on vertical transfers in
461	the atmospheric boundary layer. Boundary-layer meteorology, 97(3), 361–383.
462	doi: 10.1023/A:1002641728075
463	Mapes, B., Milliff, R., & Morzel, J. (2009). Composite life cycle of maritime trop-
464	ical mesoscale convective systems in scatterometer and microwave satellite
465	observations. Journal of the Atmospheric Sciences, $66(1)$, 199 - 208. doi:
466	https://doi.org/10.1175/2008JAS2746.1
467	Schmit, T. J., Gunshor, M. M., Menzel, W. P., Gurka, J. J., Li, J., & Bachmeier,
468	A. S. (2005). Introducing the next-generation advanced baseline imager on
469	goes-r. Bulletin of the American Meteorological Society, 86(8), 1079 - 1096.
470	doi: https://doi.org/10.1175/BAMS-86-8-1079
471	Sikora, T. D., Young, G. S., Shirer, H. N., & Chapman, R. D. (1997). Estimating
472	convective atmospheric boundary layer depth from microwave radar imagery
473	of the sea surface. Journal of Applied Meteorology, 36(7), 833 - 845. doi:
474	nttps://doi.org/10.1175/1520-0450(1997)036(0833:ECABLD)2.0.CO;2
475	Stevens, B., Bony, S., Farrell, D., Ament, F., Blyth, A., Fairall, C., Zöger, M.
476	(2021). EUREC ⁴ A. Earth System Science Data Discussions, 2021, 1–78. doi:
477	10.5194/essd-2021-18
478	Stopa, J. E., Wang, C., Vandemark, D., Foster, R., Mouche, A., & Chapron, B.
479	(2022). Automated global classification of surface layer stratification using
480	high-resolution sea surface roughness measurements by satellite synthetic
481	aperture radar. Geophysical Research Letters, $49(12)$, e2022GL098686. doi: https://doi.org/10.1020/2022GL008686
482	$\frac{1}{10000000000000000000000000000000000$
483	10rri, G., & Kuang, Z. (2019). Un cold pool collisions in tropical boundary lay-
484	ers. Geophysical Research Letters, 40(1), 399-407. doi: https://doi.org/10 1020/2018CI 080501
485	1025/20100L000001 Torri C. Kuong 7. Iz Tion V. (2015). Machanisma for convection triggering by
486	cold pools <i>Combusical Research Letters</i> 10(6) 1042 1050 doi: https://doi
487	org/10/1002/2015CL/0632227
488	.018/10.1002/201001003221

489	Vandemark, D., Mourad, P. D., Bailey, S. A., Crawford, T. L., Vogel, C. A., Sun, J.,
490	& Chapron, B. (2001). Measured changes in ocean surface roughness due to
491	atmospheric boundary layer rolls. Journal of Geophysical Research: Oceans,
492	106(C3), 4639-4654. doi: https://doi.org/10.1029/1999JC000051
493	Villefranque, N., Williamson, D., Couvreux, F., Hourdin, F., Gautrais, J., Fournier,
494	R., Volodina, V. (2020). Process-based climate model development har-
495	nessing machine learning: Iii. the representation of cumulus geometry and
496	their 3d radiative effects. Earth and Space Science Open Archive, 30. doi:
497	https://doi.org/10.1002/essoar.10505088.1
498	Vogel, R., Konow, H., Schulz, H., & Zuidema, P. (2021). A climatology of trade-
499	wind cumulus cold pools and their link to mesoscale cloud organization. Atmo-
500	spheric Chemistry and Physics, 21(21), 16609–16630. doi: https://doi.org/10
501	.5194/acp-21-16609-2021
502	Wang, C., Mouche, A., Foster, R. C., Vandemark, D. C., Stopa, J. E., Tandeo, P.,
503	Chapron, B. (2019). Characteristics of Marine Atmospheric Boundary
504	Layer Roll Vortices from Sentinel-1 Sar Wave Mode. IGARSS 2019 - 2019
505	IEEE International Geoscience and Remote Sensing Symposium, 7908-7911.
506	doi: https://doi.org/10.1109/IGARSS.2019.8900287
507	Wang, C., Vandemark, D., Mouche, A., Chapron, B., Li, H., & Foster, R. C. (2020).
508	An assessment of marine atmospheric boundary layer roll detection using
509	sentinel-1 sar data. Remote Sensing of Environment, 250, 112031. doi:
510	https://doi.org/10.1016/j.rse.2020.112031
511	Weckwerth, T. M., Wilson, J. W., & Wakimoto, R. M. (1996). Thermody-
512	namic variability within the convective boundary layer due to horizon-
513	tal convective rolls. Monthly Weather Review, 124(5), 769 - 784. doi:
514	$https://doi.org/10.1175/1520-0493(1996)124\langle 0769: TVWTCB\rangle 2.0.CO; 2000000000000000000000000000000000000$
515	Young, G. S., Sikora, T. D., & Winstead, N. S. (2000). Inferring marine atmo-
516	spheric boundary layer properties from spectral characteristics of satellite-
517	borne sar imagery. Monthly Weather Review, $128(5)$, $1506 - 1520$. doi:
518	10.1175/1520-0493(2000)128(1506:IMABLP)2.0.CO;2
519	Zhang, L., Shi, H., Wang, Z., Yu, H., Yin, X., & Liao, Q. (2018). Comparison of
520	wind speeds from spaceborne microwave radiometers with in situ observa-
521	tions and ecmwf data over the global ocean. Remote Sensing, $10(3)$. doi:
522	https://doi.org/10.3390/rs10030425
523	Zuidema, P., Li, Z., Hill, R. J., Bariteau, L., Rilling, B., Fairall, C., Hare, J.
524	(2012). On trade wind cumulus cold pools. Journal of the Atmospheric Sci-

⁵²⁵ ences, 69(1), 258 - 280. doi: https://doi.org/10.1175/JAS-D-11-0143.1