

Multi-model high-resolution analysis of Tropical-Like Cyclone Daniel with WRF and ICON: peculiarities and sensitivity to convection schemes.

Piero Serafini^{1,2}, Antonio Ricchi^{1,2}, Chiara Marsigli³, Cristiano D'Amico^{1,2}, Matteo Nastasi^{1,2}, Renata Pelosini⁴, and Rossella Ferretti^{1,2}

¹UNIVAQ (DSFC) - University of L'Aquila (Department of Physical and Chemical Sciences), L'Aquila, Italy

²CETEMPS - Center of Excellence in Telesensing of Environment and Model Prediction of Severe Events, L'Aquila, Italy

³ARPAE - Regional Agency for Environmental Protection in Emilia-Romagna, Bologna, Italy

⁴ARPAP - Regional Agency for Environmental Protection in Piemonte, Torino, Italy

Correspondence: Piero Serafini (piero.serafini@graduate.univaq.it)

Abstract. ~~Medicane Daniel (September 2023) featured a rapid transition from a baroclinic disturbance to a compact tropical-like vortex, challenging short-range prediction.~~ Medicane Daniel (September 2023) featured a multi-stage transition from a baroclinic disturbance to a compact tropical-like vortex, culminating in a rapid barotropic alignment on 9 September, challenging short-range prediction. This study delivers a side-by-side, high-resolution (~ 2 km) assessment of Daniel using two state of the art weather forecasting models, WRF and ICON, configured to be as comparable as possible in terms of domain, forcing and vertical discretizations. Seven numerical simulations are compared assessing also sensitivity to the convection scheme: fully explicit, deep-cumulus parameterized and independent shallow-convection options (plus ICON's grayzone setting). Analysis methods include an objective cyclone tracker that combines mean sea-level pressure and lower tropospheric geopotential structure, intensity metrics (central pressure and 10 m wind) along the track, precipitation anomalies regridded against IMERG observations (Integrated Multi-satellitE Retrievals for GPM). Tropical characteristics are examined with Hart's Cyclone Phase Space and Temporal Annular Symmetric Mean (TASM) of equivalent potential temperature and wind to distill three-dimensional, time-mean storm structure during the peak warm-core phase.

Both models reproduce Daniel's life cycle and produce realistic tracks. Intensity of the cyclone sharply varies from simulation to simulation, with different behavior of each model at changes in convection scheme.

The study emphasizes the different responses of the two models both in reproducing such an extreme meteorological phenomenon and in the variation of the convection scheme. Practical suggestions are established depending on the case study and the resolution used.

Copyright statement. This work is distributed under the Creative Commons Attribution 4.0 License.

1 Introduction

20 The Mediterranean Sea frequently experiences cyclones that either originate within the basin or migrate from surrounding regions. Some systems undergo rapid intensification and may even develop tropical-like characteristics, producing strong winds, heavy rainfall and coastal flooding, which can lead to significant regional impacts (Carniel et al., 2024).

Formally $\ll$ a “MEDiterranean hurriCANE” (medicane) is a mesoscale cyclone that develops over the Mediterranean Sea and displays tropical-like cyclone characteristics: a warm core extending into the upper troposphere, an eye-like feature in its center
25 with spiral cloud bands around, an almost windless center surrounded by nearly-symmetric sea-surface wind circulation with maximum wind speed within a few tens of km from the center $\gg$ (Miglietta et al., 2025). Hence, the terms medicanes and Mediterranean Tropical-Like Cyclones (TLCs) should be considered equivalent.

Although formation mechanisms are partially understood (Miglietta, 2019), the catastrophic effects resulting from strong winds and heavy rainfall underscore the importance of correctly forecasting these phenomena, for early warning, prevention, and
30 adaptation (Khodayar et al., 2025).

Several studies have shown that the predictability of intense cyclones in the Mediterranean area is very low (Doiteau et al., 2024). Furthermore, there is no consensus on which model performs best in simulating TLCs, as results can vary depending on the specific event, model configuration and evaluation metrics used. Many studies have already performed sensitivity analysis on the microphysics parameterizations, the Planetary Boundary Layer schemes and on the initialization time or dataset (Ricchi
35 et al., 2017; Pytharoulis et al., 2018; Ricchi et al., 2019), showing extreme variability of the results, especially on the track of the medicanes. Among the many options available, the cumulus parameterization can play a key role, not only in terms of the path and intensity of the cyclone (Biswas et al., 2014; Miglietta et al., 2015; Saraceni et al., 2023), but also for the amount and location of the precipitation.

From this perspective, we propose to evaluate the performance of two consolidated Numerical Weather Prediction (NWP)
40 models: WRF (Weather Research and Forecasting) and ICON (ICOsahedral Nonhydrostatic).

WRF is a mesoscale model developed by the National Center for Atmospheric Research (NCAR), the National Oceanic and Atmospheric Administration (NOAA), and the Air Force Weather Agency (AFWA). It is widely used thanks to its flexibility, scalability and a comprehensive set of physical parameterizations (Skamarock et al., 2021).

ICON is a model developed by the Deutscher Wetterdienst (DWD) and the Max Planck Institute for Meteorology (MPI-M). It
45 employs a unique grid structure based on an icosahedral mesh, which allows efficient parallelization and improved representation of atmospheric processes at various scales from global to limited area (Zängl et al., 2015).

The choice of these two models is motivated by their widespread use in both research and operational forecasting. Specifically, a horizontal grid spacing of ≈ 2 km is representative of the current convection-permitting operational standard adopted by major national meteorological services and regional research institutes like ICON-D2 at ~ 2.2 km by (DWD, 2026), ICON-2I
50 at ~ 2.2 km by (ItaliaMeteo, 2026), ICON-CH2 at ~ 2.2 km by (MeteoSwiss, 2026), WRF at ~ 4 km – 1 km by (MeteoGalicia, 2026), WRF at ~ 3 km – 1 km by (CETEMPS, 2026), WRF at ~ 2 km by (CMCC, 2026).

The selected case study focuses on Medicane Daniel developed in September 2023. The event was marked by exceptionally

low forecast predictability (Flaounas et al., 2025) and led to widespread and severe socio-economic consequences, causing extensive damage and loss in large parts of Greece and Libya (Diakakis et al., 2025).

One of the objectives of this study is to perform a systematic comparison between the WRF and ICON models in their ability to simulate the track, intensity evolution, precipitation structure and tropical-like characteristics of TLC Daniel. Particular attention is paid to evaluate how consistently each model reproduces the observed dynamical and thermodynamical features associated with the system, including its warm-core structure, the mesoscale organization and the moisture distribution. Furthermore, the study aims to quantify the sensitivity of both modeling frameworks to different representations of convective processes, specifically contrasting simulations employing explicit convection, partially parameterized schemes and fully parameterized approaches. By examining these configurations, we seek to identify the extent to which convective treatment influences the storm development, precipitation and track predictability. Ultimately, providing insights into model-dependent uncertainties relevant for the simulation of Mediterranean tropical-like cyclones.

The study is structured as follows: Section 2 provides insights into the selected case study. Section 3 describes the data and methods used, including model configurations, observational data and evaluation metrics. Section 4 presents the results of the simulations, focusing on track, intensity, precipitation and tropical characteristics. Section 5 discusses the findings, highlights the strengths and weaknesses of each model and provides conclusions and recommendations for future ~~researche~~ research.

2 Case study: Tropical-Like Cyclone Daniel

The TLC Daniel developed on 4 September 2023 and lasted until 12 September 2023 and still represents one of the most destructive Mediterranean cyclones on record. Its life-cycle can be subdivided into five main phases: genesis, extratropical development, transition, tropical-like maturity and dissipation (Fig.1 (data from ERA5 (Hersbach et al., 2023))). The system caused catastrophic flooding in Greece during its initial stages and later in Libya at peak intensity during the tropical-like phase, resulting in severe socioeconomic impacts and loss of life (Hewson et al., 2024).

Daniel originated on 4 September 2023 over the Aegean Sea as an extratropical disturbance embedded within a pronounced omega blocking pattern over Europe (Fig.1(a)). The anticyclonic Rossby wave breaking resulted in the intrusion of an upper-level PV streamer into the central Mediterranean basin, triggering cyclogenesis near the coast of Greece (Fig.1(b)).

On 5–6 September, the low-pressure system, centered in southwestern Greece, induced persistent easterly to northeasterly flow over mainland Greece. This circulation pattern facilitated strong low-level moisture transport from the Aegean and Black Seas. As a result, widespread thunderstorms developed, with cloud tops exceeding 13 km, producing extreme precipitation and severe flooding throughout central and eastern Greece, as well as parts of Bulgaria and Turkey. Surface stations recorded more than 750 mm of daily rainfall and up to 1235 mm in 4 days in the eastern parts of the Thessaly region (Lagouvardos et al., 2017).

By 7–8 September, during a slow southward drift toward the central Mediterranean, the cyclone moved over progressively warmer waters, diabatic processes intensified, marking the onset of a subtropical transition. The ~~Sea-surface-temperatures~~ Sea Surface Temperature (SST) in the south of Greece and near the Libyan coast approached 28°C, providing an anomalously large

energy reservoir for the system (Argüeso et al., 2024).

On 9 September, Daniel underwent a marked structural transformation. The surface pressure continued to decrease, the winds intensified and the vortex barotropically aligned (Fig.1(c)(d)). These characteristics are consistent with a tropical-like cyclone, or medicane.

- 90 Daniel made landfall near Benghazi at approximately 01:00 UTC on 10 September 2023. At peak intensity, the cyclone produced strong winds up to 70–80 ~~km/h~~ km/h and exceptional accumulated rainfall in northeastern Libya. Observed precipitation ranged from 150 to 240 mm in many locations: Al-Bayda recorded a daily accumulation of 414.1 mm, establishing a new national rainfall record. The extremely large runoff led to the collapse of two ageing dams near Derna, resulting in catastrophic flooding that destroyed entire neighbourhoods and caused thousands of fatalities (IOM, 2023; WMO, 2023).
- 95 After the landfall, Daniel began to weaken as it moved eastward on land, reducing its marine energy source. By 11 September, the cyclonic circulation persisted at low levels in the east of Libya and the north of Egypt. During this phase, counter-clockwise flow around the system facilitated the uploading and advection of mineral dust from the ~~western-Egyptian-desert~~ Western Desert of Egypt, generating a significant dust outbreak observed in satellite imagery. Dust transport continued until September 12 as the system further dissipated.
- 100 In summary, the impact of Daniel can be divided into two distinct stages: the cyclogenesis phase that impacts Greece and the mature tropical-like phase that impacts Libya. Forecast skill differed markedly between these stages. Medium-range ensemble predictions exhibited limited skill in forecasting cyclogenesis beyond lead times of 4–5 days, mainly due to difficulties in accurately representing the PV streamer intrusion. In contrast, once the cyclone had formed, the ensemble forecasts showed greater confidence in predicting its track and eventual landfall in Libya (Flaounas et al., 2025).

105 3 Data and Methods

3.1 Models configuration

The NWP models used for the simulations are WRF v4.6.0 and ICON v2.6.6 in Limited Area Model (LAM) configuration. In order to ease the comparison, the differences between the two models are reduced by setting the running configurations as close as possible.

- 110 Only one domain is used for both models, the WRF one was built first. The Lambertian conformal conical projection facilitates the creation of a domain whose contour does not intersect high mountain ranges along the edges or especially at the corners, avoiding possible disturbances that could result from the orography. Then the ICON domain in a rectangular plan was then adapted as best as possible to cover the same area as WRF (Fig. 2(a)). The resulting domains extend approximately in the range [11.3° E, 31.6° E, 28.5° N, 43.1° N].

115

The spatial resolution was set to 2 km, because operationally this is approximately the resolution used by both models. Consequently, in compliance with the Courant-Friedrichs-Lewy condition, the timestep was set to 6 s for both models. The runs are initialized at 00:00 UTC on September 4, 2023, and end at 00:00 UTC on September 12, 2023, with output every hour. The

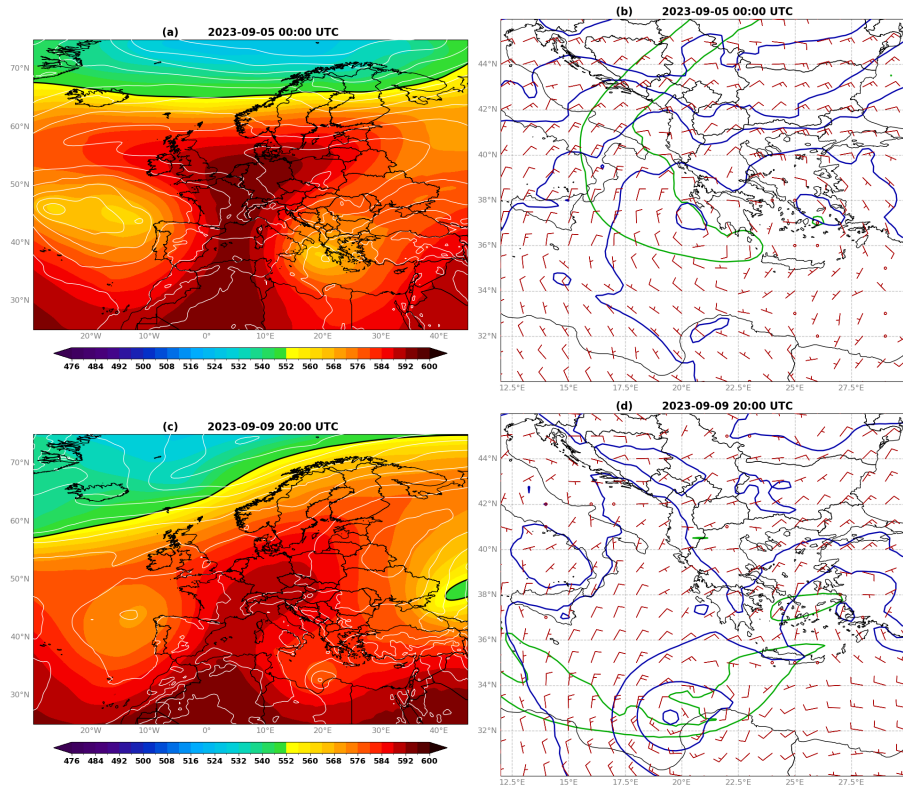


Figure 1. (a) Synoptic scale structure of TLC Daniel at: genesis stage 2023-09-05 00:00 UTC for 500 hPa Geopotential (gpdm, filled contours) and for MSLP (hPa, white contours); (b) 300 hPa 2 PVU (green contour), MSLP (blue contour) and 850 hPa wind (red barbs). At the tropical stage 2023-09-09 20:00 UTC: (c) the same as (a) and (d) the same as (b). Data from ERA5.

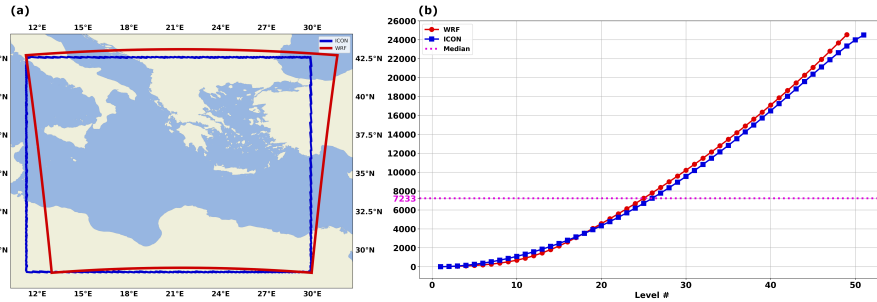


Figure 2. (a) The domains outline for WRF (red) and ICON (blue) models. (b) The vertical levels distribution for WRF (red) and ICON (blue) models; in purple dashed line the height of the median of the levels distribution.

analyses produced with the Integrated Forecasting System (IFS) by the European Centre for Medium-Range Weather Fore-
120 casts (ECMWF) (ECMWF, 2023) are used as Initial (IC) and Lateral Boundary Conditions (LBC), with Boundary Conditions

every 6 hours with a nominal spatial resolution of $0.8^\circ \times 0.8^\circ$. In order to validate the long simulation carried out for Daniel, three separate simulations were performed, initialized according to the three major phases described, covering the entire event with approximately 72 hours each. But the outcome is a deterioration in the reproduction of the track and tropical characteristics, highlighting how the re-initialization of variables, such as hydrometeors, causes imbalances in the energy budget, with a consequent reduction in the models' ability to develop a TLC probably due to the spin-up (cold start) of each re-initialised run; a dedicated analysis is left for future work.

The choice of a single domain with a grid spacing of ≈ 2 km was carefully evaluated. Given the ≈ 9 km LBCs, a direct downscaling to 1 km would introduce severe boundary reflections, while intermediate nests introduce cascading interpolation artifacts and complex dynamical noise (Miglietta et al., 2023). Furthermore, differing structural nesting ratio constraints between the models (e.g., odd integers in WRF versus powers of 2 in ICON) would produce unmatched parent domains, compromising a clean model intercomparison. Physically, ≈ 2 km represents an optimal intermediate “sweet spot” between coarse parameterizations and fully resolved ones (Brumer et al., 2026). The grid-spacing regime, ≈ 2 km is classified as convection-permitting rather than a classic deep-convection gray zone (typically 3–10 km); nevertheless, it sits squarely in the physical gray zone for shallow convection, boundary-layer turbulence, and individual updraft dynamics, which remain only partially resolved.

Vertical discretization is one of the important characteristics of the model to be taken into account when configuring it. In fact, a good representation of the lower layers is necessary where the cyclone interacts with the sea surface, as well as a good representation of the upper layers, where the jet stream and upper tropospheric disturbances are located. Both models use the hybrid coordinate: WRF uses a hybrid sigma-pressure coordinate, while ICON uses a hybrid height-based Gal-Chen coordinate, each of which requires specific parameters that are difficult to match. To obtain a distribution of vertical levels as similar as possible between the two models, an algorithm was developed in Python, that includes the translation of the internal code of the models, obtaining an output like in Fig.2(b). The algorithm acts as follows:

1. Set the parameters in common to the two models: number of vertical levels, height in meters of the lowest layer, maximum distance in m between two adjacent levels;
2. Set the height in Pascals of the highest layer for WRF and find all possible distributions varying the stretching parameters for the lower and upper layers in the range [1.0, 2.0];
3. Select the WRF configuration having half of the layers (i.e. the median) as close as possible to 5500 m (mid-troposphere);
4. Automatically set ICON parameters according to WRF distribution;
5. Find all possible ICON distributions varying the stretching factor in the range [0.0, 2.0];
6. Select ICON distribution closer to the selected one for WRF;
7. Correct the choice on WRF to improve the median matching.

Hence, the main variables are set as follows:

- number of vertical levels = 50,
- height of the lowest layer = 20 m,
- 155 – maximum distance between two adjacent levels = 1000 m,
- height of the highest layer = 5000 Pa – 24500 m .

The chosen physical parameterizations are summarized in Table 1.

Despite the effort to balance the configurations for the two simulations as much as possible, the physical parameterization suites of WRF and ICON could not be fully harmonized, reflecting the independent development lineages of the two modelling systems (Table 1). For each physics category we retained each model’s standard, operationally validated configuration rather than forcing an artificial correspondence that neither code is designed to support. The most consequential divergence concerns cloud microphysics: the WDM6 scheme (Lim and Hong, 2010) predicts number concentration only for the warm-rain species, retaining a single-moment treatment for the ice-phase hydrometeors, whereas the two-moment scheme (Seifert and Beheng, 2006) used in ICON prognoses both mass and number concentration across all hydrometeor categories, including the ice phase. Radiative transfer is likewise computed by conceptually distinct codes, with RRTMG (Iacono et al., 2008) following a correlated-k approach, in contrast to the more modular, multi-solver architecture of ecRad (Hogan and Bozzo, 2018). Boundary-layer turbulence differs in closure order: the non-local, first-order YSU scheme (Hong et al., 2006), paired with a first-order Smagorinsky closure for residual 3D mixing in WRF, compared to ICON’s prognostic, 1.5-order TKE-based scheme (Raschendorfer, 2001). Surface exchanges are computed by the single-tile Unified Noah land-surface model in WRF (Niu et al., 2011), as opposed to the explicitly tiled TERRA scheme in ICON (Schrodin and Heise, 2001), which resolves sub-grid land-cover heterogeneity relevant to the complex coastlines of the Mediterranean basin. Gravity-wave drag is included in both configurations, though following distinct orographic formulations (Choi and Hong, 2015; Orr et al., 2010). Finally, convection is parameterized by mass-flux schemes of different heritage: the KIM Simplified Arakawa–Schubert scheme in WRF (Han and Pan, 2011; Kwon and Hong, 2017), versus the Tiedtke–Bechtold scheme in ICON (Tiedtke, 1989; Bechtold et al., 2008), the latter treating shallow and deep convection as explicitly separate regimes. These structural differences, intrinsic to the two model architectures, represent an additional, model-dependent source of uncertainty that should be considered alongside the dynamical and resolution-related aspects of the comparison discussed above.

In addition, we are also testing the sensitivity to the convective schemes because the two models have differences in the choice of configuration. The following numerical experiments are performed:

- 180 – WRF explicit, i.e. no active convection parameterization (WRF–EXP),
- WRF fully parameterized calculation, both deep and shallow convection (WRF–CU),
- WRF independent shallow cumulus option WRF–SH (not tied to deep convection and GRIMS shallow cumulus from YSU group)
- ICON explicit, i.e. no active convection parameterization (ICON–EXP),

Table 1. Settings for the physical parameterizations for WRF and ICON models

Parametrization	WRF	ICON
Microphysics	WDM 6-class scheme	Double-moment
	(mp_physics = 16)	(inwp_gscp = 4)
	Lim and Hong (2010)	Seifert and Beheng (2006)
Radiation	RRTMG	ecRad
	(ra_physics = 4)	(inwp_radiation = 4)
	Iacono et al. (2008)	Hogan and Bozzo (2018)
Boundary Layer + Turbulence	YSU scheme + ff Smagorinsky 1 st closure	Prognostic TKE
	(bl_pbl_physics = 1) + (km_opt = 4)	(inwp_turb = 1)
	Hong et al. (2006)	Raschendorfer (2001)
Surface	Unified Noah land-surface model	Tiled TERRA
	(sf_surface_physics = 2)	(inwp_surface = 1)
	Niu et al. (2011)	Schrodin and Heise (2001)
Gravity Wave Drag	(gwd_opt = 1)	(inwp_gwd = 1)
	Choi and Hong (2015)	Orr et al. (2010)
Cumulus	KIM Simplified Arakawa-Schubert (KSAS)	Mass-flux shallow and deep
	(cu_physics = 14)	(inwp_convection = 1)
	Han and Pan (2011); Kwon and Hong (2017)	Tiedtke (1989); Bechtold et al. (2008)

- 185
- ICON fully parameterized calculation, both deep and shallow convection (ICON–CU),
 - ICON shallow cumulus option tied to deep convection (ICON–SH)
 - ICON grayzone parametrization tied to deep convection, (ICON–GZ).

The grayzone parametrization [used in ICON–GZ](#) is a NWP-specific tuning of the deep convection scheme in order to reduce the activity of the convection scheme to just a bit more than pure shallow convection. It includes an increased entrainment

190

rate for parcel ascent calculations and a modified CAPE closure in order to suppress widespread convective drizzle (Deutscher Wetterdienst, 2025).

A summary of the numerical experiments is in Table 2.

Table 2. Numerical experiments performed using WRF and ICON using several configurations

ACRONYM	FULL NAME	CONVECTION PARAMETERS
WRF-EXP	WRF fully explicit	cu_physics=0
WRF-SH	WRF with shallow cumulus	cu_physics=0 and shcu_physics=3
WRF-CU	WRF fully parameterized	cu_physics=14
ICON-EXP	ICON fully explicit	inwp_convection=0
ICON-SH	ICON with shallow cumulus	inwp_convection=1 and lshallowconv_only=.TRUE.
ICON-GZ	ICON grayzone	inwp_convection=1 and lgrayzone_deepconv=.TRUE.
ICON-CU	ICON fully parameterized	inwp_convection=1

3.2 Observations

195 Observation plays a key role in validating the results of numerical simulations, providing the necessary benchmark to assess
their accuracy and realism. However, in meteorological events such as TLC Daniel, which evolved primarily over offshore re-
gions, the spatial density and temporal frequency of available observations are markedly scarce. Therefore, any comprehensive
validation effort is challenging. During the initial extra-tropical phase of the event, a reasonable number of surface pressure
and wind measurements were available from several land-based meteorological stations, as well as from ships operating off the
200 coast of Greece. Hence, a relatively reliable data set is available for the early development of the system. In contrast, during
the following phase over the Ionian Sea, observational coverage became largely sparse. The situation is even more critical in
Libya, where the number of official weather stations is limited and they are far from the area directly affected by the cyclone.
These data were retrieved from official sources and used to estimate the sea-level pressure and wind at ground as mentioned
in (Hérincs, 2023) and are drawn on the map to highlight the position of the data and the displacement from the center of the
205 cyclone (Fig. 3(a)).

In the open sea, intermittent satellite-based wind observations were available from the Advanced Scatterometer (ASCAT)
instruments onboard the MetOp-B and MetOp-C satellites, operated by the European Organization for the Exploitation of Me-
teorological Satellites (EUMETSAT). These measurements are valuable for estimating surface wind fields on the sea side, but
it is well known that ASCAT tends to underestimate wind speeds in regions affected by deep convection, as is typical in the
210 case of tropical or tropical-like systems (Miglietta et al., 2025)(Fig. 3(b)).

~~Because of the dramatic lack of ground-based observations, additional insight into the storm’s structure and intensity was
obtained through satellite-derived precipitation estimates.~~ During the genesis and extra-tropical phases, the dense National
Observatory of Athens automatic network (NOANN/METEO) (Lagouvardos et al., 2017) provides high-resolution surface and
rainfall observations. However, offshore over the open Ionian Sea and in Libya, there is a dramatic lack of ground-based obser-

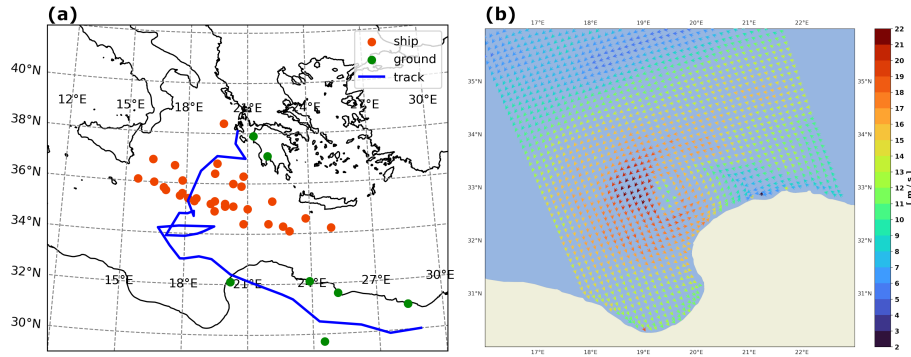


Figure 3. Available observations used for the estimation Central Sea Level Pressure and wind speed (Hérincs, 2023). (a) Location of ships, ground and buoys stations. In blue the track of the cyclone, in green the land-based stations, in orange ships and buoys. (b) The ASCAT MetOp-C wind speed measurements, at average time 20:02 UTC on September 9, 2023, wind arrows for direction in color shade according to wind speed.

215 variations. For the sake of consistency, we have decided to use satellite-derived precipitation estimates for the evaluation over the whole domain. During extreme weather events, Satellite-based rainfall products are recorded with sufficient accuracy, not only the amount of rainfall but also its spatial distribution (Kolios and Papavasileiou, 2024; Katsanos et al., 2024). In particular, data from the Integrated Multi-satellitE Retrievals for Global precipitation measurement (IMERG) (Huffman et al., 2014) product are used to evaluate the two-dimensional cumulative precipitation fields, offering a more complete representation of the atmospheric conditions during the peak of the event. The dataset refers to the “final” run v07 satellite-gauge product merged into half-hourly $0.1^\circ \times 0.1^\circ$ (roughly 10x10 km nominal spatial resolution) fields.

220 IMERG Final incorporates gauge information only through a monthly climatological adjustment and underestimated the observed maxima over Thessaly during this event (Flaounas et al., 2025); model overestimates diagnosed over Greece against IMERG should hence be read as upper bounds.

225 3.3 Cyclone track

One of the interesting information for a cyclone is definitely the track, that is the path that the cyclone traces during its life. Therefore, an analysis of it is performed: first, the observed track is built and used as reference; then, the simulated one is computed based on the model results; finally, the distance between these two is computed.

3.3.1 Observed track

230 The observed track is obtained through the High Resolution Visible (HRV) images produced by the Spinning Enhanced Visible Infra-Red Imager (SEVIRI) Meteosat Second-Generation (MSG) satellite and provided by EUMETSAT. A Python algorithm is used to manually select the location of the center of the cyclone. For each available image, the algorithm acts as a graphical interface, showing the image and allowing the user to select the point where the pressure minimum is assumed to be located.

This selection is made by selecting the center of the eye, where present, or the point of maximum convergence of the convective system. The longitude, latitude, and date are extracted from the selection. The temporal resolution of the track is set to 3 hours and a nominal spatial resolution of 1 km, but with an uncertainty of about 15 km due to the difficulty in identifying the exact position of the cyclone center.

3.3.2 Tracking algorithm

For what concerns the tracks of the model simulations, an ‘ad hoc’ algorithm is newly developed. Basically, the tracking technique consists of following the minimum mean sea level pressure. However, this can cause errors. The high-resolution simulations for deep convection events may develop a few deep pressure minima located beneath the most intense cells, which forces the algorithm to deviate by the deepest minimum. To overcome this issue, our algorithm is built by choosing the centroid of the cells with values of mean sea level pressure lower than the 5th percentile in a radius of 150 km from the previously detected cyclone center. To ensure the track toward the deepest minimum, this constraint is also applied to the geopotential levels between 900 and 800 hPa, to allow for good tracking even in a baroclinic atmosphere. The final result is an average of the individual minimum and centroid points. To ease the visual representation of the cyclone track a smoothing is applied using a 3-hour moving mean window.

In order to evaluate the goodness of the simulated tracks, we use the mean error (μ), the standard deviation (σ) and the Root Mean Squared Error (RMSE) computed on the geodetic distance w.r.t. the observed track. [Since the uncertainty on the observed track applies equally to all simulations and through the track, it does not affect the error evaluation of the configurations, but it is still taken into account while evaluating the statistical significance of the results.](#)

3.4 Meteorological variables

To analyse this event, several models meteorological variables are used:

- Sea Level Pressure (SLP), both Mean SLP (MSLP) and Central SLP (CSLP), i.e. minimum MSLP at the center of the cyclone, which provide an indication of cyclone intensity ; [the observational data are drawn from \(Hérincs, 2023\), which relies on records from ships, buoys, and official weather stations, with the underlying trend then reconstructed by means of the UKMet analysis.](#)
- Wind speed, at 10 m above ground, which can be compared with the Saffir-Simpson Hurricane Wind Scale (SSHWS) to estimate cyclone strength ; [the observational data are drawn from \(Hérincs, 2023\), which relies on records from ships, official weather stations and satellite data.](#)
- Precipitations patterns, accumulated amount and location

The following parameters are analysed along the cyclone track as follows: time series are built for the CSLP and the 10 m maximum wind (MW10); the minimum pressure and the maximum wind are extracted at each timestep within a radius of 150 km of the cyclone center and plotted separately highlighting landfall and absolute extremes. Moreover, the simulated total

265 accumulated precipitation over the whole duration of the event is compared with the IMERG one.

Model-minus-IMERG difference maps of the event-total precipitation, computed on the common verification grid (Fig. A4), show that the leading error structure along the track is a displacement dipole rather than a uniform intensity bias. WRF overestimates accumulation over the Pindus slopes, ICON underestimates it on the southern flank of the cyclone, and the dipole amplitudes are smallest in the SH configurations of both models.

270 The Fractions Skill Score (FSS) is used to evaluate rainfall. Both observed and model rainfall are regridded in a common spatial domain. The FSS is defined as (Roberts and Lean, 2008):

$$FSS_{\tau} = 1 - \frac{\sum_w (p_{w,\tau}^O - p_{w,\tau}^S)^2}{\sum_w (p_{w,\tau}^O)^2 + \sum_w (p_{w,\tau}^S)^2} \quad (1)$$

where τ is a fixed threshold, w is the dimension of the window, $p_{w,\tau}^{O/S}$ number of Observed/Simulated grid cells in the window w whose accumulated precipitation exceeds the value τ and $\sum_w$ is the sum over all window sizes ranging from 1 cell to 171
275 cells (maximum extension equal to almost the entire domain). The following thresholds are defined:

- 6th percentile = 1 mm
- 50th percentile = 16 mm
- 85th percentile = 108 mm
- 95th percentile = 193 mm
- 280 – 99th percentile = 308 mm
- 99.9th percentile = 428 mm

~~Hence, a perfect forecast is represented by the FSS approaching 1; low values represent a poor forecast. To be pointed out, the FSS strongly depends on the size of the window, if high values for FSS are obtained for small window, then the model performs well. However, high FSS values associated with large windows suggest a poor localization of the precipitation but a good~~
285 ~~accumulation value.~~ The FSS evaluates the spatial overlap of fractional coverage between model and observations, ranging from 0 (complete mismatch) to 1 (perfect forecast). Because the score is symmetric, a low FSS value indicates a general mismatch in fractional coverage, which can stem equally from spatial displacement, over-forecasting, or under-forecasting. To rigorously assess the spatial scale at which a model becomes skilful, we employ the “useful scale” criterion introduced by (Roberts and Lean, 2008). A forecast is considered skilful when the FSS exceeds a target value defined as $FSS_{useful} =$
290 $0.5 + f_0/2$, where f_0 is the domain-average fractional coverage of the observed event. Since f_0 is inherently very small for localized extreme precipitation (such as the upper percentiles analysed here), the target threshold is effectively $FSS_{useful} \approx 0.5$. The minimum window size required to reach this value represents the spatial scale at which the model provides a reliable precipitation forecast.

3.5 Physical processes of the event

295 3.5.1 Cyclone Phase Space Diagram (CPS)

The analysis of the physical mechanisms of the phenomenon is focused on the tropical characteristics of the cyclone. A kind of cyclone identity card can be created by exploiting the three-dimensional Cyclone Phase Space Diagram (CPS) developed by Hart et al. (Hart, 2003).

300 The three parameters used to describe the general structure of cyclones are the lower-tropospheric thermal asymmetry (B) and the thermal wind (VTL), and the upper-tropospheric thermal wind (VTU). The first phase space diagram generated is the storm-motion-relative thickness asymmetry (B vs VTL) identifying areas symmetric/non-frontal versus asymmetric/frontal. The second is the vertical derivative of the horizontal height gradient (VTU vs VTL) that describes the structure of cold-versus warm-core via the thermal wind relationship. The parameters are calculated according to (Hart, 2003) using values in a circular area around the center of the cyclone with a radius $R = 300$ km.

305 ~~Medicane-adapted CPS variants employ smaller radii and shifted layers (e.g., 100 km and 925–700–400 hPa (Picornell et al., 2014; Miglietta et al., 2025)). A sensitivity test on ICON–SH, varying the radius between 100 and 500 km and testing both layer sets, shows that the numerical values of B , $-V_T^L$ and $-V_T^U$ fluctuate between setups while the effect sizes remain small (mean Cohen’s d between -0.28 and 0.35), and that the adopted configuration (300 km, standard layers) lies near the centre of the tested ensemble (mean absolute z-score 0.24). The original Hart framework is retained for comparability with the existing~~
310 ~~literature on tropical and tropical-like cyclones. The coloured thresholds in Figs. 8-9 are indicative, qualitatively inspired by the SSHWS, and serve only as a visual aid for comparing simulations.~~

3.5.2 Temporal Annular Symmetric Mean (TASM)

The equivalent potential temperature (θ_e) is used to explore the internal structure of the cyclone during the tropical phase. This variable is used because it accounts for both dry potential temperature (sensible heating) and amount of water vapor (latent
315 heating) and it is conserved along the process, making it a good tracer for humid air-masses and easily comparable at different heights.

~~The cyclone structure is typically analysed using vertical cross-sections of θ_e .~~ Vertical cross-sections of θ_e are often employed to inspect the moist thermodynamic and warm-core structure of tropical and tropical-like cyclones (Emanuel, 2018; Miglietta and Rotunno, 2019). This approach is easy to use for tropical cyclones, where the symmetric structure is well defined and the
320 required spatial resolution is not too high. However, when applied to TLCs in the Mediterranean, it is highly dependent on the direction of the cutting for the cross-section since symmetry is not so defined and the need for high-resolution simulations makes the field even more variable. For this reason, we used what we will call “Temporal Annular Symmetric Mean” (TASM). Assuming a symmetric cyclone structure, the horizontal field is divided into concentric circular coronas, centered on the cyclone’s core and of thickness ΔR . Both θ_e and the wind speed are averaged within each corona for all levels, thus generating
325 the Annular Symmetric Mean (ASM). To overcome the temporal dependence, the ASM is repeated for at least 8 consecutive

time steps, including the absolute minimum of CSLP. This results in a TASM for a good indicative tropical-like period. The results are evaluated according to the known structure of Tropical Cyclones (Emanuel, 2018; Tous, 2011).

4 Results

4.1 Track

330 At first, the model tracks of the cyclone are compared with the observed one. They are shown in Figure 4 with the following color code: WRF-EXP in red, WRF-SH in orange, WRF-CU in yellow; ICON-EXP in blue, ~~ICON-SH in petrol-green, ICON-GZ in cyan and ICON-CU in blue-gray~~ ~~ICON-SH in green, ICON-GZ in magenta, ICON-CU in light-blue~~. The observed track is the black line. Overall, all simulations reproduce the general path of the cyclone, which moves from the Aegean Ionian Sea to Libya and then turns eastward towards Egypt. However, there are significant differences in terms of spatial and
335 temporal accuracy.

To objectively evaluate the goodness of the models track the mean error and RMSE are computed using as reference the observed HRV track. The WRF-EXP simulation shows the best performance, with a mean error of 3363 km and a RMSE of 3976 km Fig.4(c). The cyclone is accurately tracked throughout its life cycle, with a slight eastward bias during the landfall phase. The WRF-SH simulation has a mean error of 4074 km and a RMSE of 5096 km. The cyclone is correctly tracked in the early
340 phase, but it deviates significantly during the landfall phase, with a westward bias. The WRF-CU simulation has a mean error of 4576 km and a RMSE of 5586 km. The cyclone is poorly tracked in the early phase, with a northward bias, but the track improves during the landfall phase, with a slight eastward bias.

The ICON-EXP simulation has a mean error of 3898 km and a RMSE of 45110 km. The cyclone is well tracked in the early phase, but it deviates significantly from the observed track during the landfall phase, with a westward bias. The ICON-SH
345 simulation has a mean error of 4283 km and a RMSE of 4897 km. The cyclone is poorly tracked in the early phase, with a northward bias, but improves during the landfall phase, with a slight eastward bias. The ICON-GZ simulation has a mean error of 50105 km and a RMSE of 60129 km. The cyclone is poorly tracked throughout its life cycle, with a significant northward bias. The ICON-CU simulation has a mean error of 5592 km and a RMSE of 65102 km. Again, the cyclone is poorly tracked in the early phase, with a northward bias, but it improves during the landfall phase, with a slight eastward bias.

350 ~~In general, WRF performs slightly better than ICON in simulating the track of TLC Daniel~~ While WRF exhibited a lower mean track error compared to ICON, rigorous significance testing accounting for serial correlation demonstrates that this difference is not statistically significant ($p = 0.28$), being largely overshadowed by the substantial variability along the cyclone life cycle. In particular, WRF better reproduces the barotropic conditions (TL phase) than the baroclinic ones, whereas ICON seems to be better in the baroclinic storm phase 4(b). The choice of cumulus parameterization also affects the performance ~~, with explicit and shallow cumulus schemes generally yielding better results than fully parameterized schemes~~. Although explicit and shallow cumulus schemes yielded lower absolute mean errors in our sample compared to fully parameterized schemes, these improvements are statistically indistinguishable once serial correlation is accounted for. Therefore, a strict ranking of

individual configurations based solely on track error cannot be robustly supported for this specific event.

360 A detailed analysis of the landfall phase shows that ICON–SH has the smallest spatial error, but a time delay of 10 hours. The best time synchronization is provided by ICON–GZ, but with a spatial error of 35 km. The best overall simulation appears to be WRF–SH, which has a spatial error of only 9 km and a delay of 5 hours. The other simulations show larger displacements, ranging from 35 to 122 km in space and from 2 to 15 hours in time. A summary of this analysis is shown in Table 3.

Table 3. Summary of landfall data including location, date, displacements in space and in time [w.r.t. satellite observation of landfall](#).

Sim	Lat	Lon	Date	ds [km]	dt [h]
OBS	32.43	20.43	2023–09–10 01:00	~	~
WRF–EXP	32.63	20.84	2023–09–10 08:00	45.0	7.0
WRF–SH	32.47	20.51	2023–09–10 06:00	9.0	5.0
WRF–CU	32.36	20.45	2023–09–10 08:00	8.0	7.0
ICON–EXP	31.98	20.02	2023–09–10 10:00	63.0	9.0
ICON–SH	32.45	20.46	2023–09–10 11:00	3.0	10.0
ICON–GZ	32.16	20.24	2023–09–10 03:00	35.0	2.0
ICON–CU	32.83	21.64	2023–09–10 16:00	122.0	15.0

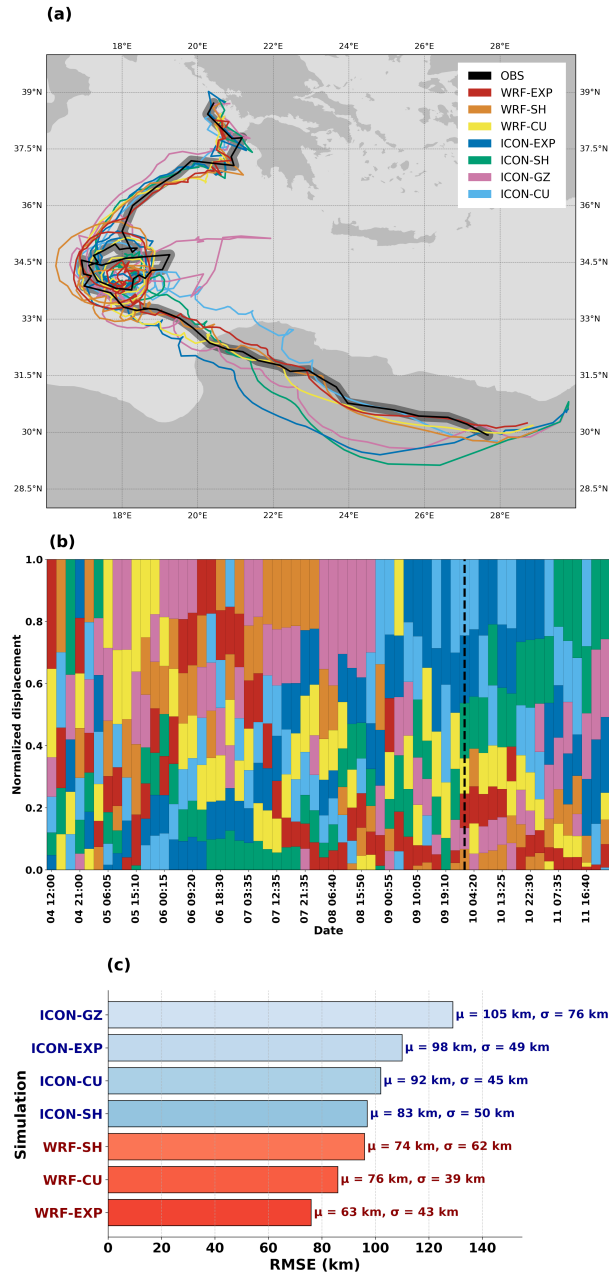


Figure 4. (a) The cyclone tracks for: WRF-EXP in red, WRF-SH in orange, WRF-CU in yellow; ICON-EXP in blue, ~~ICON-SH in~~ petrol-green, ~~ICON-GZ in~~ cyan, ~~ICON-CU in~~ blue-grey, observed track is the black line. ~~ICON-SH in~~ green, ~~ICON-GZ in~~ magenta, ~~ICON-CU in~~ light-blue, observed track is the black line with a shadowed buffer indicating the estimated 15 km error. (b) Normalized error in km w.r.t. the observed track; bars are stacked vertically for each timestep, i.e. lower is better; same colors of panel a. (c) RMSE, mean value and standard deviation of each simulation.

4.2 Meteorological variables

365 The comparison among the models meteorological variables and the observed one is now presented. The time series of CSLP and MW10 along the cyclone track are shown in Figures 5(a) and 5(b) respectively. The observed time series (OBS) are retrieved from (Hérincs, 2023). All simulations manage to reproduce the general trend of the variables, with a decrease in CSLP and an increase in MW10 during the most intense phase of the cyclone. However, there are significant differences in terms of intensity and timing.

370 To be noticed, the observed minimum CSLP of 996 hPa may be an overestimation due to the lack of data in the area where the cyclone reached its peak intensity. In fact, the Benina International Airport, which is the closest available observation to the cyclone, recorded a surface pressure of 984 hPa at an elevation of approximately 132 m above sea level and reported a sea level pressure of 999.3 hPa at 00:00 UTC on 10 September 2023, more than 50 km away from the cyclone's center. Similar considerations can be made for the wind speed. The ASCAT Metop-C passing over the cyclone at 20:02 UTC on 9 September 375 2023, shows a maximum wind speed of 22 ms^{-1} , but it is known that scatterometer measurements tend to underestimate the real value in situations of deep convection (Miglietta et al., 2025).

As a reference of strength of the cyclone, according to the SSHWS, an hurricane can be classified in category C1 if the maximum sustained wind speed is between 33 and 42 ms^{-1} . While according to the observed value from ASCAT it should be categorized as a Tropical Storm.

380 During the most intense phase of the cyclone, significant differences are found among the model simulations in terms of the minimum central sea level pressure (CSLP) (Fig. 5(a)) and the maximum 10-m wind speed (MW10) (Fig. 5(b)). For what concerns the WRF simulations: WRF-EXP produced the lowest minimum CSLP of 982 hPa, with MW10 approximately 36.2 ms^{-1} around the cyclone center during its maximum intensity; WRF-SH yielded a higher minimum CSLP of 990 hPa, with MW10 of 31.8 ms^{-1} , while the WRF-CU configuration produced a minimum CSLP of 983 hPa, with MW10 of 35.6 ms^{-1} .

385 Similarly, for the ICON simulations: ICON-EXP produced a minimum CSLP of 993 hPa and MW10 of 27 ms^{-1} ; ICON-SH yielded 986 hPa with MW10 of 25.4 ms^{-1} ; ICON-CU reached a minimum CSLP of 980 hPa, accompanied by MW10 of 32.8 ms^{-1} , representing the most intense results among the ICON configurations. In contrast, the ICON-GZ setup exhibited the weakest system, with a minimum CSLP of 1001 hPa and MW10 of 24.2 ms^{-1} .

Overall, the WRF configurations tend to produce lower CSLP values and larger MW10 speeds compared to the ICON ones, 390 indicating a generally more intense cyclone representation. While the shallow parameterization for WRF decreases both the intensity and the strength of the cyclone, on the contrary for ICON it tends to allow a greater deepening in terms of CSLP while decreasing MW10. Fully parameterized configurations increase both the intensity and strength of the cyclone, likely by creating deep convection cells that rotate in phase around an axis. This is particularly true for ICON, where ICON-CU reaches the lowest CSLP and the strongest MW10; while in WRF the WRF-CU experiment reaches intensities similar to those of 395 WRF-EXP.

The observed CSLP minimum is used only as a qualitative reference for the temporal evolution; the quantitative statements in this section only compare the simulations with one another.

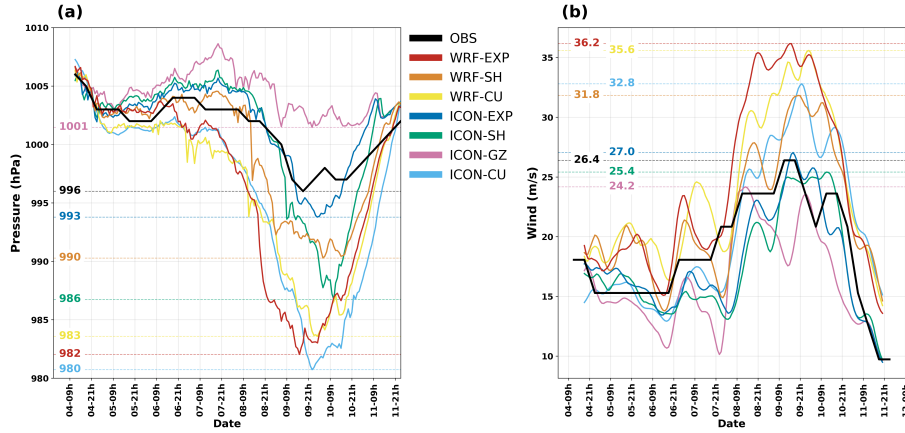


Figure 5. Timeseries of the cyclone variables: (a) CSLP every 6h for observation and 1h for the simulations; (b) Maximum wind speed within a radius of ~~100~~150 km from cyclone center every 6 h for observation and 1 h for simulations using a 6 h running mean. WRF-EXP in red, WRF-SH in orange, WRF-CU in yellow; ICON-EXP in blue, ICON-SH in green, ICON-GZ in magenta, ICON-CU in light-blue. Reference behaviour of the observations in black

The total accumulated precipitation during the whole event (4-12 September 2023) is shown in Figure 6. All simulations are able to reproduce the general pattern of the precipitation, with a maximum over the Ionian Sea and the northwestern Greece.

400 However, there are significant differences in terms of quantity and location.

The mean accumulated precipitation per grid cell is an indicator of the total rainfall in the domain. This quantity has been computed as the spatial average of the event-total accumulated precipitation over the full analysis domain, on the common verification grid shared by both models and IMERG. The following values are obtained: 31.4 mm for IMERG, 34.8 mm for WRF-EXP, 33.3 mm for WRF-SH, 34.3 mm for WRF-CU, 29.3 mm for ICON-EXP, 27.0 mm for ICON-SH, 39.1 mm for ICON-CU, 23.9 mm for ICON-GZ. These values show that the precipitation is systematically larger for the WRF experiments, while the ICON simulations are generally drier, although ICON-CU produces the largest accumulated rainfall value and ICON-GZ is markedly drier than the other configurations.

The spatial distribution of the precipitation shows that WRF produces (Fig.6(b)(c)(d)) a pattern consistent with observations, highlighting the three circular structures related to the different precipitation phases. On the other hand, ICON-SH shows (Fig. 6 (g)) a remarkable improvement with respect to the other ICON configurations (Fig. 6(e)(f)(h)).

~~The FSS is now used for a quantitative analysis of the simulations. The FSS highlights the peculiarities of each simulation. If the size of the window is small and the FSS is high, the model agree well with the observations. As the window size increases, so does the number of grid cells that exceed the accumulation threshold; if FSS is close to one, the number of cells between simulation and observation is similar; if FSS is lower, the simulation has more cells than the observations. At low thresholds, all models have good scores (Fig. 7 (a)(b)(c)(d)(e)(f)(g) black and blue lines). The first differences are noticeable from the 85th percentile, where the best results are achieved by the fully parameterized ICON-CU scheme (Fig. 7 (g) green line). On the contrary, rising to the 95th percentile threshold, the best result for WRF is in the SH configuration, but ICON-EXP performs~~

415

better at small window (Fig. 7 (b) and (e) yellow line respectively). The 99th and 99.9th thresholds show the most dramatic differences among the experiments: WRF-SH is the only one achieving $FSS > 0$ on the 1×1 grid at both a threshold of 99% and 99.9% (Fig. 7 b red and violet lines), indicating a good localization of extreme precipitation accumulations. Conversely, ICON-CU yields the lowest FSS at the same threshold (Fig. 7 (g) violet line), indicating that there are much more cells exceeding the threshold. ICON-SH and ICON-EXP show a sharp increases and the highest FSS values at larger window size (Fig. 7 (e) and (f), red line), resulting in low displacement error and good amount of precipitation for extremes. On the overall, ICON-EXP attains the highest FSS values. The FSS is now used for a quantitative analysis of the simulated precipitation fields.

Assessing the FSS as a function of the neighbourhood window size highlights the specific spatial capabilities of each simulation. High FSS values at small window sizes indicate an accurate spatial localization of the precipitation. Conversely, when the FSS only rises at larger window sizes, it implies that the model captures the overall precipitation frequency but suffers from significant displacement errors at the grid scale. At low thresholds, all models exhibit high skill scores, rapidly exceeding the $FSS_{useful} \approx 0.5$ criterion even at small spatial scales (Fig. 7a-d, black and blue lines). Noticeable differences emerge starting from the 85th percentile, where the fully parameterized ICON-CU scheme achieves the best performance at finer scales (Fig. 7g, green line). However, as the threshold rises to the 95th percentile, the best grid-scale result for WRF is obtained with the SH configuration, while ICON-EXP demonstrates the highest skill among the ICON simulations at small window sizes (Fig. 7b and 7e, yellow line). The 99th and 99.9th percentiles highlight the most dramatic discrepancies among the experiments. WRF-SH is the only configuration achieving an $FSS > 0$ at the native grid scale (1×1) for both extreme thresholds (Fig. 7b, red and violet lines), indicating some exact spatial overlap for extreme local accumulations. Conversely, ICON-CU yields the lowest overall FSS at these extremes (Fig. 7g, violet line), reflecting a severe mismatch in fractional coverage. ICON-SH and ICON-EXP, while exhibiting displacement errors at the native grid scale, show a sharp increase in skill at larger window sizes, ultimately attaining the highest overall FSS values (Fig. 7e and 7f, red lines). This indicates that although these setups struggle to pinpoint the exact location of extreme cells, they successfully reproduce the correct fractional coverage of extreme precipitation over broader areas, reliably crossing the FSS_{useful} threshold at synoptic scales. Overall, taking both peak accumulations and spatial scales into account, ICON-EXP attains the most robust FSS performance.

Examining the minimum skillful ($FSS \geq 0.5$) window sizes across configurations and thresholds reveals that nearly all simulations achieve skill at the grid scale (1 grid-cell) up to the 85th percentile, whereas none reach skillfulness at the 99.9th percentile threshold. The main differences emerge at the 99th percentile threshold, where ICON-EXP achieves skillfulness at the smallest window size (19 grid-cells), followed by WRF-CU (41 grid-cells), WRF-EXP and WRF-SH (49 grid-cells), ICON-SH (67 grid-cells), and ICON-GZ (85 grid-cells). ICON-CU fails to reach an FSS threshold of 0.5. The frequency distributions of event-total precipitation, evaluated on the common verification grid (Fig. A5-A6), show agreement with FSS scores.

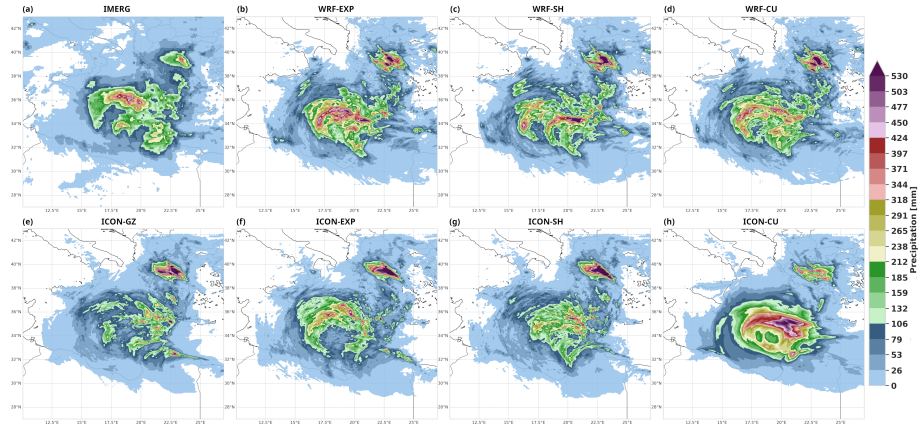


Figure 6. Accumulated total precipitations during the event (4-12 September 2023). (a) Observations from IMERG, (b) WRF-EXP, (c) WRF-SH, (d) WRF-CU, (e) ICON-GZ, (f) ICON-EXP, (g) ICON-SH, (h) ICON-CU. Color shades as for FSS percentile thresholds in Fig. 7.

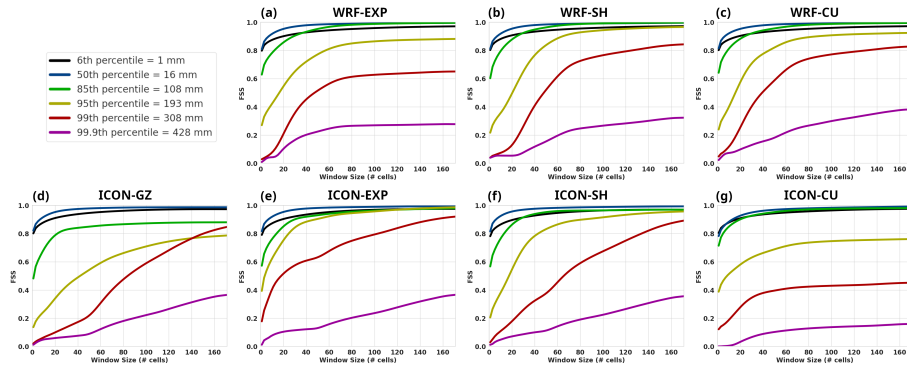


Figure 7. FSS scores for each simulation with respect to IMERG. On the x-axis the window size used for the FSS computation. Color shades based on percentiles consistent with Fig. 6.

4.3 Physics Physical structure of the cyclone

450 In this section, the tropical characteristics of the cyclone are analysed to the aim of understanding its physical structure. The CPS diagrams for the lower symmetry (B vs VTL plane) and core characteristics (VTU vs VTL plane) are shown respectively in Figures 8 and 9. All simulations show a general similar behaviour in the B vs VTL plane, with an initial phase of decreasing asymmetry and increasing warm core until reaching a maximum value of VTL. During this tropical-like phase the symmetry tends to oscillate around zero, keeping the structure in the expected range.

455 But differences are found for the two models: WRF shows (Fig. 8(a)(b)(c)) a gradual increase in core temperature (VTL) while oscillating in symmetry until the dissipation stage, which is characterized by a rapid decrease of the core temperature and an increase in the asymmetry. On the other hand, ICON tends to warm up the core later than WRF but more rapidly (Fig. 8(d)(e)(f)). Moreover, as expected from the track analyses, ICON-GZ does not develop a Tropical-Like Cyclone, keeping a cold core and chaotic disorganized structure (Fig. 8(g)).

460 Also for the VTU vs VTL plane, all simulations show a similar behaviour, with an initial phase of contemporaneous warming in the upper and lower levels until reaching a peak in upper right section, followed by a phase of cold and shallow core. However, there are significant differences in terms of balance in the lower and upper warming. WRF-SH and ICON-EXP seem to have difficulties in transferring energy to the upper layers resulting in a colder core in the VTU axis (Fig. 9(b) and (d), respectively). As previously pointed out ICON-GZ does not even reach the threshold to enter a tropical-like classification (Fig. 9 (g)).

465 To better understand the structure of the cyclone produced by the two models, a further analysis is performed using the Temporal Annular Symmetric Mean which allows for exploring the internal structure of the cyclone during the tropical phase. The typical structure of a vertical cross-section of θ_e for a Tropical Cyclone reveals a pronounced warm column all around the center extending from the lower troposphere to approximately 200 hPa. Usually at the 500 hPa there is a relative minimum due to subsidence and reduced moisture that inhibit deep convection creating the eye of the cyclone. Around it, larger values

470 of θ_e form a sort of pillars representing the vigorous ascent and strong diabatic heating of the eye wall. Beyond it, θ_e decreases steadily with distance from the center (Emanuel, 2018).

The simulations of both models show maximum values of θ_e in the lower levels, close to the cyclone center, increasing up to 800 hPa and gradually decreasing horizontally away from it (Fig. 10). This indicates strong diabatic heating in the lower troposphere near the minimum SLP and the extraction of heat from the sea. However, there are significant differences among

475 the simulations in terms of vertical extent and horizontal distribution.

WRF-EXP and WRF-CU show (Fig. 10(a) and (c), respectively) a well-defined warm core extending up to approximately 300 hPa, with a clear eye wall structure and a relative decrease of θ_e in the range of 650-600 hPa and a horizontal extension of a few tens of km. The wind contours are packed near the eyewall reaching maximum values at 50-100 km far from the center at 900 hPa. Similarly for WRF-SH (Fig. 10(b)) but producing a less pronounced eyewall and a relative minimum of θ_e

480 around 550 hPa. The wind contours are less packed than for WRF-EXP, but they are still following the eyewall, the maximum values can be found at 100-150 km far from the center at 900 hPa.

ICON-EXP shows (Fig. 10(d)) a warm core extending up to 550 hPa, with poor connection with the upper layers. Wind con-

tours are packed approaching the cyclone center with maximum values 50-100 km from the center below 900 hPa. ICON-SH shows (Fig. 10(e)) a warm core extending up to 300 hPa, with an eyewall extremely close to the center and a relative minimum of θ_e around 625 hPa, but not aligned with the central axis. The wind contours are packed in the first ~~tenth~~ tens of kilometers from the center with maximum values at 50-100 km at 900 hPa. ICON-CU shows (Fig. 10(f)) values of θ_e reaching larger values than the other ICON simulations, with similar characteristics, except for a deep intrusion of θ_e from the upper layers, suggesting the presence of intense cells that rotate organized around a minimum pressure, but without the development of a real tropical-like cyclone. ICON-GZ shows (Fig. 10(g)) a very weak cyclone structure with a thick layer of low θ_e values from 650 to 400 hPa and wind contours distant from each other with almost half the values compared to other simulations. These results suggest that the ICON simulations are more sensitive to the cumulus scheme used than the WRF ones.

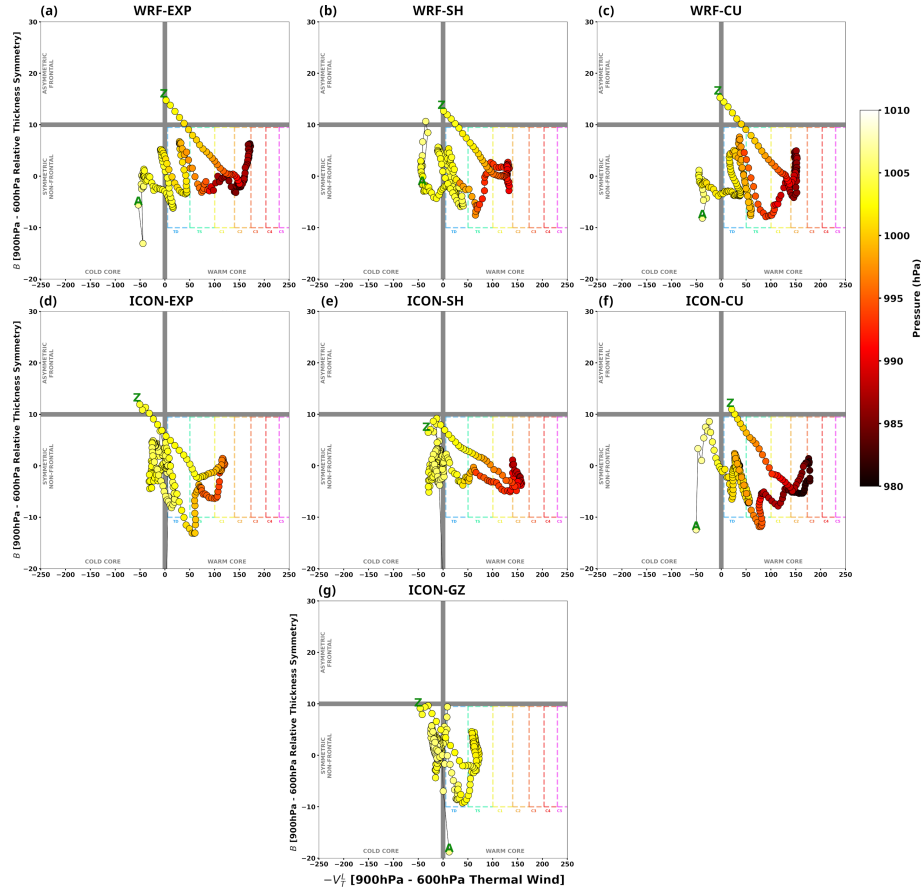


Figure 8. CPS B vs VTL plane: core lower tropospheric symmetry. The letters A to Z indicate the cyclone life cycle, with time moving forward along the trajectory, from the labelled A to Z. The shading of the markers corresponds to the intensity of the cyclone (in hPa), with light color as the weakest and dark color as the most intense. Rainbow dashed lines indicate the thresholds for the cyclone phase classification.

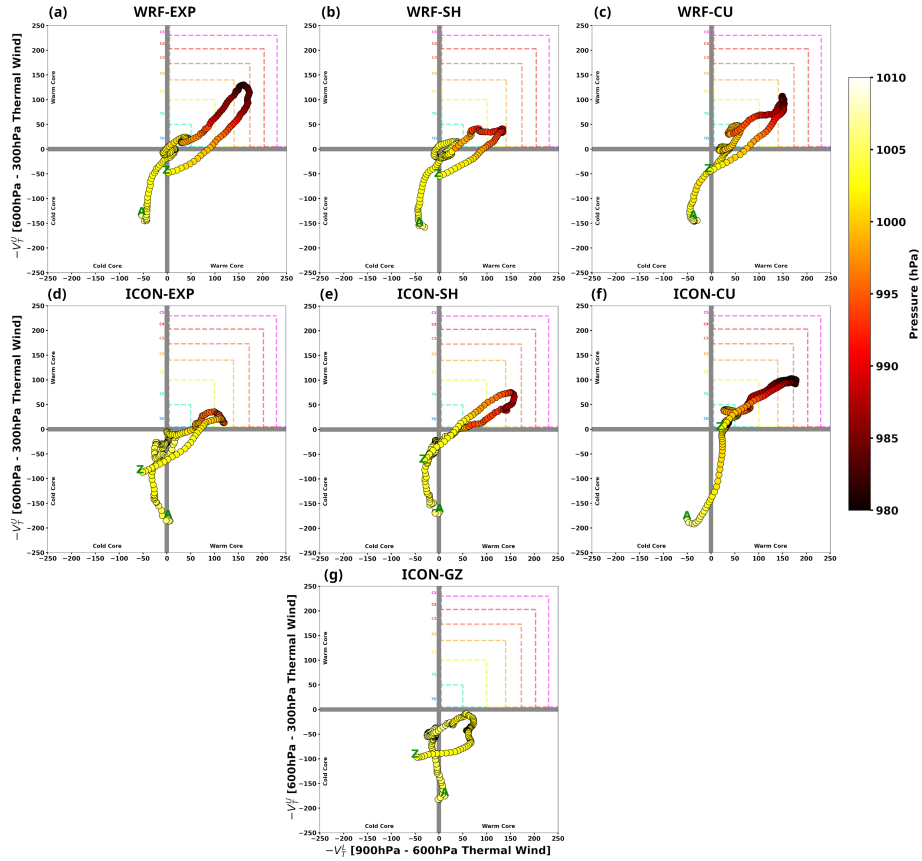


Figure 9. CPS VTU vs VTL plane: core temperature and deepness. The letters A to Z indicate the cyclone life cycle, with time moving forward along the trajectory, from the labelled A to Z. The shading of the markers corresponds to the intensity of the cyclone (in hPa), with light color as the weakest and dark color as the most intense. Rainbow dashed lines indicate the thresholds for the cyclone phase classification.

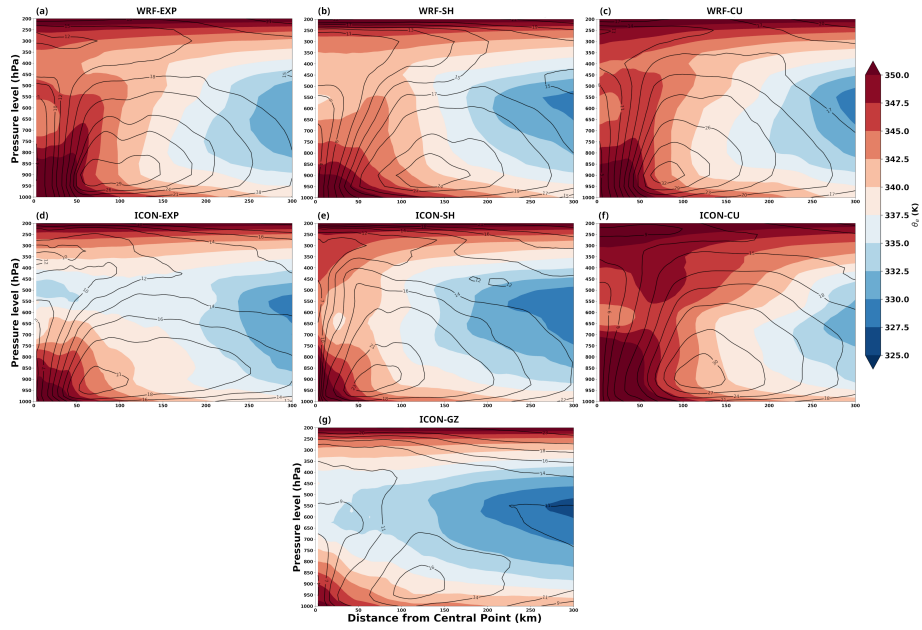


Figure 10. TASM: wind speed (m/s, black contours) and equivalent potential temperature (K, filled contours) for the simulations using different cumulus parameterizations. Each panel is averaged over 8 timesteps before landfall. (a) WRF-EXP from 10-Sep-2023 01:00 UTC to 10-Sep-2023 08:00 UTC. (b) WRF-SH from 09-Sep-2023 23:00 UTC to 10-Sep-2023 06:00 UTC. (c) WRF-CU from 10-Sep-2023 01:00 UTC to 10-Sep-2023 08:00 UTC. (d) ICON-EXP from 10-Sep-2023 03:00 UTC to 10-Sep-2023 10:00 UTC. (e) ICON-SH from 10-Sep-2023 04:00 UTC to 10-Sep-2023 11:00 UTC. (f) ICON-CU from 10-Sep-2023 09:00 UTC to 10-Sep-2023 16:00 UTC. (g) ICON-GZ from 09-Sep-2023 20:00 UTC to 10-Sep-2023 03:00 UTC.

5 Discussion and Conclusion

In this study, an inter-comparison of high-resolution simulations of Medicane Daniel between two leading weather forecasting models, WRF and ICON, is presented. The objective is to understand the role of convective parameterization at the ≈ 2 km “grayzone” resolution and the different representations of the event by the two models. In order to ensure a robust comparison, the following model settings were constrained to be similar: domain, vertical discretizations, initial and lateral boundary conditions (IC/LBC) and physical parameterizations. Despite the effort to balance the configurations for the two simulations as much as possible, the physical parameterization suites of WRF and ICON could not be fully harmonized, reflecting the independent development lineages of the two modelling system. These structural differences, intrinsic to the two model architectures, represent an additional, model-dependent source of uncertainty.

The experiments reveal that the transition from a baroclinic disturbance to a tropical-like cyclone (TLC) is primarily driven by sub-grid scale convection. Although both models capture the broad life cycle of Daniel, significant differences in wind intensity, surface pressure, accumulated precipitation and internal structure highlight the strong control exerted by physical parameterizations, even within convection-permitting regimes.

Both models show differences in the storm structure that vary during the stages of the storm. ICON demonstrates good accuracy during the initial extratropical phase, likely benefiting from its global-model heritage and efficient representation of large-scale baroclinic dynamics. In contrast, WRF exhibits a reduced track error and greater consistency among configurations during the mature tropical-like phase, suggesting that its mesoscale-focused physics may be better tuned for the diabatic processes dominating TLC Daniel maintenance in these simulations.

The inclusion of a shallow convection (SH) parameterization in this experiment proved to be important for both models, significantly improving landfall accuracy near Benghazi compared to fully explicit (EXP) or fully parametrized (CU) configurations. This suggests that at ≈ 2 km resolution, the explicit computation of shallow non-precipitating clouds is insufficient at least for the Daniel case. Instead, parameterizing these clouds is fundamental to modulate boundary layer moisture flux and preconditioning the environment for organized deep convection, thereby refining the storm’s trajectory.

The sensitivity to the shallow-convection treatment admits a consistent physical reading. At 2 km grid spacing, non-precipitating shallow cumuli are not resolved. The shallow schemes (GRIMS in WRF, the shallow-only Tiedtke–Bechtold mode in ICON) vent boundary-layer moisture into the lower free troposphere; the sub-cloud layer dries, grid-scale moisture convergence weakens and the onset of resolved deep convection is delayed, so that latent heating is released over a wider area rather than in a few grid-point updrafts. This mechanism is consistent with the weaker but better organised cyclone in WRF–SH, with the improved localisation of extreme rainfall (Fig. 7(b)) and with the smoother eyewall structure in the TASM diagnostics (Fig. 10). In the CU experiments, instead, the parameterized deep heating acts on top of the resolved convection and feeds a positive feedback between heating and central pressure fall, consistent with the over-intensification of ICON–CU. This interpretation is diagnostic; a quantitative attribution would require moisture-budget and heating-profile analyses beyond the scope of this study.

Numerical simulations reveal a systematic bias regarding the intensity of the cyclone: WRF consistently produces deeper central sea-level pressures (CSLP) and stronger maximum 10 m winds (MW10) than ICON. Physically, this is explained by the

different vertical structures identified in the Temporal Annular Symmetric Mean (TASM) diagnostics. The WRF configurations generate a vertically coherent warm core that extends up to 300 hPa, with a distinct eyewall-like θ_e maximum. This vertical alignment favours barotropic intensification and reduces tilt-induced ventilation, enhancing hydrostatic pressure falls through the thermal wind relationship. In contrast, ICON-EXP exhibits a shallow warm anomaly, largely confined below 550 hPa, which limits the depth-integrated thickness anomaly and subsequent surface pressure deepening. The ICON-GZ (grayzone) experiment represents a critical failure mode, as it over-damps convective development and prevents the transition to a diabatic, Wind-Induced Surface Heat Exchange (WISHE)-like maintenance mechanism. This suggests that, at this resolution, the GZ parameterization suppresses the updraft generation too much without compensating for the missing energy with advective flow, inhibiting the development of deep convection necessary for the tropical transition. Moreover, an over-intensification of the cyclone is obtained by WRF-CU and ICON-CU simulations, which schemes trigger an unrealistic positive feedback loop between latent heat release and pressure deepening, leading to extreme CSLP values. However, the internal structure of the cyclone differs; while WRF-CU maintains a concentrated eye-wall structure, ICON-CU exhibits a broader moisture distribution and lacks a distinct structural “pillar”, suggesting that while deep-convection schemes can produce intense pressure drops, they may struggle to realistically organize inner-core dynamics at high resolution.

These structural discrepancies are mirrored in the precipitation analysis. WRF tends to over-produce precipitation, while ICON under-produces it with respect to IMERG data. The Fraction Skill Score (FSS) analysis indicates that WRF-SH uniquely attains skill on the grid scale for extreme rainfall thresholds. This suggests that a hybrid approach, allowing explicit deep convection while retaining shallow parameterization, enhances the localization of convective cores without over-smoothing updrafts. Overall, the EXP and SH configurations retain the largest FSS values for both models, particularly ICON-EXP.

The results must be interpreted in light of Daniel’s intrinsic peculiarity as a long-lived, multi-stage Mediterranean cyclone. Unlike systems that rapidly attain a mature structure, Daniel evolved through a prolonged sequence of physically distinct stages: an initial baroclinic phase, a tropical transition and a mature TLC phase. This temporal persistence across heterogeneous regimes makes the event a stringent benchmark for testing model physics, requiring a single framework to represent large-scale forcing and mesoscale convective organization with comparable skill.

Several factors limit the generalization of these findings. First, the lack of high-frequency, in-situ surface observations over the open Mediterranean necessitates reliance on satellite-derived tracks and reanalysis for validation. Second, as a single-case design, these results may be specific to Daniel’s unique dynamical pathway, especially given the high variability reported in the literature for other medicanes. Furthermore, the role of microphysics interacting with convection and air-sea feedback (e.g., sea spray and wave-induced drag) remains an unquantified source of uncertainty that may account for some intensity differences.

The present findings sit consistently within the medicane sensitivity literature. The first-order control of the convective treatment on track and intensity confirms the single-case results of (Miglietta et al., 2015), and the benefit of a dedicated shallow-convection treatment at kilometre scale agrees with (Saraceni et al., 2023). The large spread among configurations, with no single setup performing best across all phases and metrics, mirrors the multi-physics results of (Ricchi et al., 2017, 2019; Pytharoulis et al., 2018). Daniel departs from most previously studied events in its prolonged, multi-stage evolution, which stresses model physics across heterogeneous regimes within a single integration. Accordingly, the operational indications

given below are to be read as case-specific: findings established for Daniel under the present configurations, findings consistent with prior studies, and hypotheses that remain to be tested across additional cases and initialization times.

For operational purposes, these results suggest that ≈ 2 km “convection-permitting” simulations should not be run as purely explicit. A hybrid approach using shallow convection schemes acts as a “bridge”, allowing models to resolve organized deep convection while parameterizing necessary smaller-scale mixing. However, the failure of ICON–GZ and the over-intensification of CU schemes underlines that scheme portability cannot be assumed. At this resolution, the solution is governed by the combined action of physical parameterizations and the numerical framework; therefore, the same physical option may not be dynamically equivalent across different models due to differences in discretization, diffusion and vertical coordinates. It follows that each model requires dedicated and finely tuned calibration.

Part of the systematic WRF–ICON contrast may stem from the unmatched physics rather than from the convective treatment. The single-moment ice phase in WDM6, as opposed to the fully two-moment Seifert–Beheng scheme, plausibly alters the vertical distribution of latent heating and hence the warm-core depth diagnosed in the TASM; the non-local YSU mixing, compared with ICON’s prognostic-TKE closure, modulates the boundary-layer moisture supply and the near-surface wind maxima. Cross-model statements in this study should hence be read as differences between full model configurations, whereas causal statements about the convection scheme are restricted to the within-model experiments.

The large sensitivity found in this work reinforces the need for case-specific forecast strategies. For long-lived systems like Daniel, researchers should not assume that a single “optimal” setup is robust across all phases; instead, an in-depth sensitivity analysis should be performed on a case-by-case basis. Future research should expand to multi-physics ensembles to assess the coupling between boundary layer schemes and the successful shallow-convection setups identified here. In physically heterogeneous areas such as the Mediterranean basin, model deficiencies during the transition from baroclinic to tropical-like regimes may propagate into substantial errors in rainfall distribution, amount and landfall timing. Hence, early-warning systems should not rely exclusively on nominal horizontal resolution, but also on demonstrated model skill for the specific class of high-impact events being targeted. Additionally, investigating the role of Gravity Wave Drag (GWD) in modulating the upper-level PV streamer’s interaction with the surface vortex could provide further insights into the predictability of the transition phase. Finally, high-resolution coupled atmosphere–ocean simulations are needed to determine how SST feedbacks modulate vertical warm-core depth and whether such feedbacks could mitigate the intensity biases observed in WRF and ICON.

Data availability. IMERG dataset: Huffman, G.J., E.F. Stocker, D.T. Bolvin, E.J. Nelkin, Jackson Tan (2023), GPM IMERG Final Precipitation L3 Half Hourly 0.1 degree x 0.1 degree V07, Greenbelt, MD, Goddard Earth Sciences Data and Information Services Center (GES DISC), Accessed: [16-Dec-2025], 10.5067/GPM/IMERG/3B-HH/07 .

ERA5 dataset: Hersbach, H., Bell, B., Berrisford, P., Biavati, G., Horányi, A., Muñoz Sabater, J., Nicolas, J., Peubey, C., Radu, R., Rozum, I., Schepers, D., Simmons, A., Soci, C., Dee, D., Thépaut, J.-N. (2023): ERA5 hourly data on pressure levels from 1940 to present. Copernicus Climate Change Service (C3S) Climate Data Store (CDS), DOI: 10.24381/cds.bd0915c6 (Accessed on 04-Nov-2025)

Author contributions. PS contributed to the formal analysis by applying statistical and computational techniques, conducted the investigation and data collection, developed the software and supporting algorithms, designed the methodology, prepared the visualizations, and wrote the original draft of the manuscript. AR curated and maintained the datasets for analysis and reuse, contributed to the development of the methodology, secured funding acquisition for the project and internal review of the manuscript. CM provided supervision of the research activities and performed critical internal review of the manuscript. CdA carried out validation of the results to ensure reproducibility and contributed to the internal review process. MN contributed to the development of the methodology and participated in the internal review of the manuscript. RP contributed to the internal review, providing critical revision and commentary on the manuscript. RF led the conceptualization of the research goals, administered the project, provided resources necessary for the study, and contributed to the internal review of the manuscript.

Competing interests. The authors declare that they have no conflict of interest.

Acknowledgements. Funding sources: The research described in this paper has been developed in the framework of the research project National Centre for HPC, Big Data and Quantum Computing - PNRR Project, funded by the European Union - Next Generation EU and by the Special Project “Exploiting Coupled, High-resolution modelling to simulate severe mediterranean cyclOgenesES (ECHOES)” founded by ECMWF”.

The authors acknowledge the use of AI-assisted tools during the development and refinement of post-processing codes. These systems were used to support code structuring, debugging, and optimization, as well as to suggest implementation strategies and documentation improvements. All generated outputs were carefully reviewed, validated, and adapted by the authors to ensure correctness, reproducibility, and consistency with the scientific objectives of the study. The authors take full responsibility for the final code and its scientific use.

References

- Argüeso, D., Marcos, M., and Amores, A.: Storm Daniel fueled by anomalously high sea surface temperatures in the Mediterranean, *NPJ Climate and Atmospheric Science*, 7, 307, <https://doi.org/10.1038/s41612-024-00872-2>, 2024.
- Bechtold, P., Köhler, M., Jung, T., Doblas-Reyes, F., Leutbecher, M., Rodwell, M. J., Vitart, F., and Balsamo, G.: Advances in simulating atmospheric variability with the ECMWF model: From synoptic to decadal time-scales, *Quarterly Journal of the Royal Meteorological Society: A journal of the atmospheric sciences, applied meteorology and physical oceanography*, 134, 1337–1351, <https://doi.org/10.1002/qj.289>, 2008.
- Biswas, M. K., Bernardet, L., and Dudhia, J.: Sensitivity of hurricane forecasts to cumulus parameterizations in the HWRF model, *Geophysical Research Letters*, 41, 9113–9119, <https://doi.org/10.1002/2014GL062071>, 2014.
- Brumer, S. E., Pantillon, F., Pianezze, J., Le Péru-Morvan, Y., Ragu-Fonta, M., Grand, L., Ricchi, A., Barboni, A., and Bouin, M.-N.: On the Tropical Nature of an Intense Mediterranean Cyclone in the Ocean-Atmosphere System, *Journal of Geophysical Research: Atmospheres*, 131, <https://doi.org/10.1029/2025JD046105>, 2026.
- Carniel, C. E., Ricchi, A., Ferretti, R., Curci, G., Miglietta, M. M., Reale, M., Serafini, P., Wellmeyer, E. D., Davolio, S., Zardi, D., et al.: A high-resolution climatological study of explosive cyclones in the Mediterranean region: Frequency, intensity and synoptic drivers, *Quarterly Journal of the Royal Meteorological Society*, 150, 5561–5582, <https://doi.org/10.1002/qj.4889>, 2024.
- CETEMPS: Center of Excellence in Telesensing of Environment and Model Prediction of Severe Event (CETEMPS) – Meteorological Modeling, <http://www.cetemps.aquila.infn.it/modellistica>, last accessed 22 July 2026, 2026.
- Choi, H.-J. and Hong, S.-Y.: An updated subgrid orographic parameterization for global atmospheric forecast models, *Journal of Geophysical Research: Atmospheres*, 120, 12 445–12 457, <https://doi.org/10.1002/2015JD024230>, 2015.
- CMCC: Euro-Mediterranean Center on Climate Change (CMCC) – Weather Research and Forecasting (WRF) Model, <https://www.cmcc.it/models/wrf>, last accessed 22 July 2026, 2026.
- Deutscher Wetterdienst: Working with the ICON model: ICON Tutorial 2025, page 107, Tech. rep., Deutscher Wetterdienst (DWD), https://doi.org/10.5676/DWD_pub/nwv/icon_tutorial2025, 2025.
- Diakakis, M., Sarantopoulou, A., Gogou, M., Filis, C., Nastos, P., Kapris, I., Vassilakis, E., Konsolaki, A., and Lekkas, E.: Cascade effects induced by extreme storms and floods: The case of storm daniel (2023) in greece, *Water*, 17, 912, <https://doi.org/10.3390/w17070912>, 2025.
- Doiteau, B., Pantillon, F., Plu, M., Descamps, L., and Rieutord, T.: Systematic evaluation of the predictability of different Mediterranean cyclone categories, *Weather and Climate Dynamics*, 5, 1409–1427, <https://doi.org/10.5194/wcd-5-1409-2024>, 2024.
- DWD: Deutscher Wetterdienst (DWD) – Numerical Weather Prediction Forecast Data, https://www.dwd.de/EN/ourservices/nwp_forecast_data/nwp_forecast_data.html, last accessed 22 July 2026, 2026.
- ECMWF: IFS Documentation CY48R1 - Part II: Data Assimilation, 2, ECMWF, <https://doi.org/10.21957/a744f32e74>, 2023.
- Emanuel, K.: 100 years of progress in tropical cyclone research, *Meteorological Monographs*, 59, 15–1, <https://doi.org/10.1175/AMSMONOGRAPHS-D-18-0016.1>, 2018.
- Flaounas, E., Dafis, S., Davolio, S., Faranda, D., Ferrarin, C., Hartmuth, K., Hochman, A., Koutroulis, A., Khodayar, S., Miglietta, M. M., et al.: Dynamics, predictability, impacts and climate change considerations of the catastrophic Mediterranean Storm Daniel (2023), *Weather and Climate Dynamics*, 6, 1515–1538, <https://doi.org/10.5194/wcd-6-1515-2025>, 2025.

- Han, J. and Pan, H.-L.: Revision of convection and vertical diffusion schemes in the NCEP Global Forecast System, *Weather and Forecasting*, 26, 520–533, <https://doi.org/10.1175/WAF-D-10-05038.1>, 2011.
- Hart, R. E.: A Cyclone Phase Space Derived from Thermal Wind and Thermal Asymmetry, *Monthly Weather Review*, 131, 585–616, [https://doi.org/10.1175/1520-0493\(2003\)131<0585:ACPSDF>2.0.CO;2](https://doi.org/10.1175/1520-0493(2003)131<0585:ACPSDF>2.0.CO;2), 2003.
- 650 Hérincs, D.: Mediterranean Tropical Cyclone Report: Tropical Storm Daniel, Wikipedia, https://zipipotty.hu/2023_daniel.pdf, 2023.
- Hersbach, H., Bell, B., Berrisford, P., Biavati, G., Horányi, A., Muñoz Sabater, J., Nicolas, J., Peubey, C., Radu, R., Rozum, I., et al.: ERA5 hourly data on pressure levels from 1940 to present, Copernicus Climate Change Service (C3S) Climate Data Store (CDS), <https://doi.org/10.24381/cds.bd0915c6>, 2023.
- 655 Hewson, T., Ashoor, A., Boussetta, S., Emanuel, K., Lagouvardos, K., Lavers, D., Magnusson, L., Pilloso, F., and Zsoter, E.: Mediane Daniel: An extraordinary cyclone with devastating impacts, *ECMWF newsletters*, 179, 33–47, <https://doi.org/10.21957/th3wxk861d>, 2024.
- Hogan, R. J. and Bozzo, A.: A flexible and efficient radiation scheme for the ECMWF model, *Journal of Advances in Modeling Earth Systems*, 10, 1990–2008, <https://doi.org/10.1029/2018MS001364>, 2018.
- 660 Hong, S.-Y., Noh, Y., and Dudhia, J.: A new vertical diffusion package with an explicit treatment of entrainment processes, *Monthly weather review*, 134, 2318–2341, <https://doi.org/10.1175/MWR3199.1>, 2006.
- Huffman, G. J., Bolvin, D. T., Braithwaite, D., Hsu, K., Joyce, R., and Xie, P.: Integrated multi-satellite retrievals for GPM (IMERG), *Precipitation Measurement Missions Technical documentation*, <https://doi.org/10.5067/GPM/IMERG/3B-HH/07>, 2014.
- Iacono, M. J., Delamere, J. S., Mlawer, E. J., Shephard, M. W., Clough, S. A., and Collins, W. D.: Radiative forcing by long-lived greenhouse gases: Calculations with the AER radiative transfer models, *Journal of Geophysical Research: Atmospheres*, 113, <https://doi.org/10.1029/2008JD009944>, 2008.
- 665 IOM: Libya — Storm Daniel Flash Update 8 (13 October 2023), Global Data Institute Displacement Tracking Matrix, International Organization for Migration, <https://dtm.iom.int/fr/node/30081>, 2023.
- ItaliaMeteo: National Agency for Meteorology and Climatology (ItaliaMeteo) – Short-Range Forecasts for Italy, <https://www.agenziaitaliameteo.it/meteo/previsioni/previsioni-a-breve-termine-italia/>, last accessed 22 July 2026, 2026.
- 670 Katsanos, D., Retalis, A., Kalogiros, J., Psiloglou, B. E., Roukounakis, N., and Anagnostou, M.: Performance Evaluation of Satellite Precipitation Products During Extreme Events—The Case of the Mediane Daniel in Thessaly, Greece, *Remote Sensing*, 16, 4216, <https://doi.org/10.3390/rs16224216>, 2024.
- Khodayar, S., Kushta, J., Catto, J. L., Dafis, S., Davolio, S., Ferrarin, C., Flaounas, E., Groenemeijer, P., Hatzaki, M., Hochman, A., Kotroni, V., Landa, J., Láng-Ritter, I., Lazoglou, G., Liberato, M. L. R., Miglietta, M. M., Papagiannaki, K., Patlakas, P., Stojanov, R., and Zittis, G.: Mediterranean Cyclones in a Changing Climate: A Review on Their Socio-Economic Impacts, *Reviews of Geophysics*, 63, <https://doi.org/10.1029/2024RG000853>, 2025.
- 675 Kolios, S. and Papavasileiou, N.: Daily Rainfall Patterns During Storm “Daniel” Based on Different Satellite Data, *Atmosphere*, 15, 1277, <https://doi.org/10.3390/atmos15111277>, 2024.
- 680 Kwon, Y. C. and Hong, S.-Y.: A mass-flux cumulus parameterization scheme across gray-zone resolutions, *Monthly Weather Review*, 145, 583–598, <https://doi.org/10.1175/MWR-D-16-0034.1>, 2017.
- Lagouvardos, K., Kotroni, V., Bezes, A., Koletsis, I., Kopania, T., Lykoudis, S., Mazarakis, N., Papagiannaki, K., and Vougioukas, S.: The automatic weather stations NOANN network of the National Observatory of Athens: operation and database, *Geoscience Data Journal*, 4, 4–16, <https://doi.org/10.1002/gdj3.44>, 2017.

- 685 Lim, K.-S. S. and Hong, S.-Y.: Development of an effective double-moment cloud microphysics scheme with prognostic cloud condensation nuclei (CCN) for weather and climate models, *Monthly weather review*, 138, 1587–1612, <https://doi.org/10.1175/2009MWR2968.1>, 2010.
- MeteoGalicia: Galician Meteorological Service (MeteoGalicia) – Numerical Prediction Models, <http://www.meteogalicia.gal/web/modelos-numericos>, last accessed 22 July 2026, 2026.
- MeteoSwiss: Federal Office of Meteorology and Climatology (MeteoSwiss) – ICON Numerical Weather Prediction System, <https://www.meteosvizzer.admin.ch/tempo/sistemi-di-allerta-e-previsione/sistema-di-previsione-numerica-icon.html>, last accessed 22 July 690 2026, 2026.
- Miglietta, M. M.: Mediterranean Tropical-Like Cyclones (Medicanes), *Atmosphere*, 10, 206, <https://doi.org/10.3390/atmos10040206>, 2019.
- Miglietta, M. M. and Rotunno, R.: Development mechanisms for Mediterranean tropical-like cyclones (medicanes), *Quarterly Journal of the Royal Meteorological Society*, 145, 1444–1460, <https://doi.org/10.1002/qj.3503>, 2019.
- 695 Miglietta, M. M., Mastrangelo, D., and Conte, D.: Influence of physics parameterization schemes on the simulation of a tropical-like cyclone in the Mediterranean Sea, *Atmospheric Research*, 153, 360–375, <https://doi.org/10.1016/j.atmosres.2014.09.008>, 2015.
- Miglietta, M. M., Buscemi, F., Dafis, S., Papa, A., Tiesi, A., Conte, D., Davolio, S., Flaounas, E., Levizzani, V., and Rotunno, R.: A high-impact meso-beta vortex in the Adriatic Sea, *Quarterly Journal of the Royal Meteorological Society*, 149, 637–656, <https://doi.org/10.1002/qj.4432>, 2023.
- 700 Miglietta, M. M., Flaounas, E., González-Alemán, J. J., Panegrossi, G., Gaertner, M. A., Pantillon, F., Pasquero, C., Schultz, D. M., D’adderio, L. P., Dafis, S., et al.: Defining medicanes: Bridging the knowledge gap between tropical and extratropical cyclones in the Mediterranean, *Bulletin of the American Meteorological Society*, 106, E1955–E1971, <https://doi.org/10.1175/BAMS-D-24-0289.1>, 2025.
- Niu, G.-Y., Yang, Z.-L., Mitchell, K. E., Chen, F., Ek, M. B., Barlage, M., Kumar, A., Manning, K., Niyogi, D., Rosero, E., et al.: The community Noah land surface model with multiparameterization options (Noah-MP): 1. Model description and evaluation with local- 705 scale measurements, *Journal of Geophysical Research: Atmospheres*, 116, <https://doi.org/10.1029/2010JD015140>, 2011.
- Orr, A., Bechtold, P., Scinocca, J., Ern, M., and Janiskova, M.: Improved middle atmosphere climate and forecasts in the ECMWF model through a nonorographic gravity wave drag parameterization, *Journal of Climate*, 23, 5905–5926, <https://doi.org/10.1175/2010JCLI3490.1>, 2010.
- Picornell, M. A., Campins, J., and Jansà, A.: Detection and thermal description of medicanes from numerical simulation, *Natural Hazards and Earth System Sciences*, 14, 1059–1070, <https://doi.org/10.5194/nhess-14-1059-2014>, 2014.
- Pytharoulis, I., Katsios, S., Tegoulis, I., Feidas, H., Miglietta, M. M., Matsangouras, I., and Karacostas, T.: Sensitivity of a mediterranean tropical-like cyclone to physical parameterizations, *Atmosphere*, 9, 436, <https://doi.org/10.3390/atmos9110436>, 2018.
- Raschendorfer, M.: The new turbulence parameterization of LM, News Letter No. 1, <http://www.cosmo-model.org>, 2001.
- Ricchi, A., Miglietta, M., Barbariol, F., Benetazzo, A., Bergamasco, A., Bonaldo, D., Cassardo, C., Falcieri, F., Modugno, G., Russo, A., 715 Sclavo, M., and Carniel, S.: Sensitivity of a Mediterranean Tropical-Like Cyclone to Different Model Configurations and Coupling Strategies, *Atmosphere*, 8, 92, <https://doi.org/10.3390/atmos8050092>, 2017.
- Ricchi, A., Miglietta, M. M., Bonaldo, D., Cioni, G., Rizza, U., and Carniel, S.: Multi-Physics Ensemble versus Atmosphere–Ocean Coupled Model Simulations for a Tropical-Like Cyclone in the Mediterranean Sea, *Atmosphere*, 10, 202, <https://doi.org/10.3390/atmos10040202>, 2019.
- 720 Roberts, N. M. and Lean, H. W.: Scale-selective verification of rainfall accumulations from high-resolution forecasts of convective events, *Monthly Weather Review*, 136, 78–97, <https://doi.org/10.1175/2007MWR2123.1>, 2008.

- Saraceni, M., Silvestri, L., Bechtold, P., and Bongioannini Cerlini, P.: Mediterranean tropical-like cyclone forecasts and analysis using the ECMWF ensemble forecasting system with physical parameterization perturbations, *Atmospheric Chemistry and Physics*, 23, 13 883–13 909, <https://doi.org/10.5194/acp-23-13883-2023>, 2023.
- 725 Schrodin, R. and Heise, E.: The Multi-Layer Version of the DWD Soil Model TERRA_LM, Technical Report No. 2, <http://www.cosmo-model.org>, 2001.
- Seifert, A. and Beheng, K. D.: A two-moment cloud microphysics parameterization for mixed-phase clouds. Part 1: Model description, *Meteorology and atmospheric physics*, 92, 45–66, <https://doi.org/10.1007/s00703-005-0112-4>, 2006.
- Skamarock, W., Klemp, J., Dudhia, J., Gill, D. O., Liu, Z., Berner, J., Wang, W., Powers, J. G., Duda, M. G., Barker, D., and Huang, X.-Y.:
 730 A Description of the Advanced Research WRF Model Version 4.3, NCAR Technical Notes, <https://doi.org/10.5065/1dfh-6p97>, 2021.
- Tiedtke, M.: A comprehensive mass flux scheme for cumulus parameterization in large-scale models, *Monthly weather review*, 117, 1779–1800, [https://doi.org/10.1175/1520-0493\(1989\)117<1779:ACMFSF>2.0.CO;2](https://doi.org/10.1175/1520-0493(1989)117<1779:ACMFSF>2.0.CO;2), 1989.
- Tous: Medicanes: cataloguing criteria and exploration of meteorological environments, *Tethys, Journal of Weather and Climate of the Western Mediterranean*, <https://doi.org/10.3369/tethys.2011.8.06>, 2011.
- 735 Wilks, D. S.: *Statistical methods in the atmospheric sciences*, vol. 100, Academic press, 2011.
- WMO: Storm Daniel leads to extreme rain and floods in Mediterranean, heavy loss of life in Libya, World Meteorological Organization, <https://wmo.int/media/news/storm-daniel-leads-extreme-rain-and-floods-mediterranean-heavy-loss-of-life-libya>, 2023.
- Zwiers, F. W. and Von Storch, H.: On the role of statistics in climate research, *International Journal of Climatology: A Journal of the Royal Meteorological Society*, 24, 665–680, <https://doi.org/10.1002/joc.1027>, 2004.
- 740 Zängl, G., Reinert, D., Rípodas, P., and Baldauf, M.: The ICON (ICOsahedral Non-hydrostatic) modelling framework of DWD and MPI-M: Description of the non-hydrostatic dynamical core, *Quarterly Journal of the Royal Meteorological Society*, 141, 563–579, <https://doi.org/10.1002/qj.2378>, 2015.

Appendix A: Supplementary Materials

A1 Robustness Assessment of tracks error

745 To assess the true robustness of the track-skill ranking, we have added a dedicated serial-correlation-aware significance assessment.

The method deserves a short description, as a naive test assuming independent hourly pairs would severely overstate significance. We compare configurations through paired differences of the hourly displacement errors at common valid times. Consecutive hourly errors are strongly serially correlated because position errors persist over many hours; both the standard
750 paired t -test and the Wilcoxon signed-rank test assume independent pairs, so raw p -values computed on the full hourly sample would be spuriously small.

We account for this serial correlation in two complementary ways:

- (i) Paired t -test with effective sample size: The test is applied using an effective sample size $n_{\text{eff}} = n(1 - r_1)/(1 + r_1)$, where r_1 is the lag-1 autocorrelation of the difference series (Zwiers and Von Storch, 2004; Wilks, 2011).
- 755 – (ii) Subsampled Wilcoxon signed-rank test: As a distribution-free check, the Wilcoxon test is applied to differences subsampled at 12-hourly spacing to ensure temporal independence relative to the cyclone’s evolution.

A difference is reported as significant only when both tests reject the null hypothesis at the 5% level ($\alpha = 0.05$).

The outcome of this serial-correlation-aware analysis is as follows:

- Sample Size & Autocorrelation: With $n = 60$ common hourly samples and lag-1 autocorrelations (r_1) of the difference
760 series ranging between 0.466 and 0.872, the effective sample sizes range between $n_{\text{eff}} = 4.1$ and $n_{\text{eff}} = 21.8$.
- Pooled WRF vs. ICON Families: The difference between the pooled WRF-family and pooled ICON-family series (-23.34 km) is not significant ($t = -1.224$, $p = 0.2847$ for the n_{eff} -corrected t -test; $p = 1.0000$ for the 12-hourly Wilcoxon test).
- Pairwise Comparisons: Among the pairwise comparisons, no pair was found to be statistically significant once serial
765 correlation was accounted for; the remaining pairs are not distinguishable from baseline variability.
- Observational Uncertainty Benchmark: As a secondary, phase-resolved result, the mean displacement error of every configuration exceeds the 15 km observational uncertainty outside the cyclogenesis phase (one-sample t -test with the same n_{eff} correction), which quantifies where in the life cycle the models lose the track.

A2 Precipitation per phases

770 An analysis of the rainfall divided by cyclone phase is presented. During the extratropical phase, all configurations reproduce the event fairly accurately in terms of both location and intensity (Fig.A1). The first differences appear in the transition phase: WRF–SH shows a completely de-localized maxima and ICON–CU has no cells with values above the 99th percentile (Fig.A2).

Finally, in the tropical phase, all configurations maintain a fairly good localization of the precipitation, except for ICON-CU (Fig.A3).

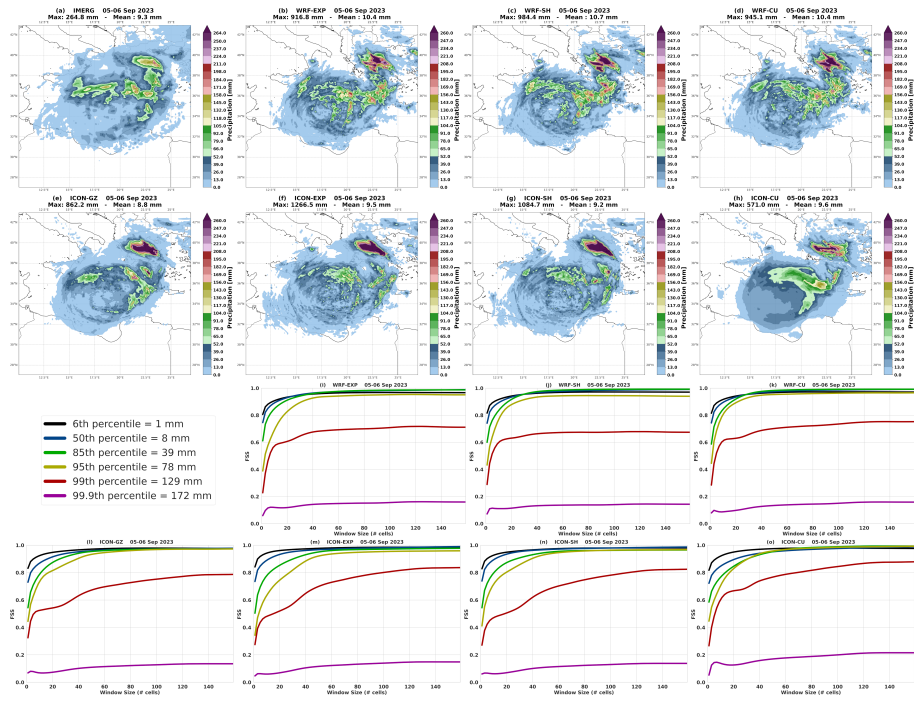


Figure A1. Accumulated total precipitations during extratropical phase (from 2023-09-05 to 2023-09-06). (a) Observations from IMERG, (b) WRF-EXP, (c) WRF-SH, (d) WRF-CU, (e) ICON-GZ, (f) ICON-EXP, (g) ICON-SH, (h) ICON-CU. Color shades as for FSS percentile thresholds. FSS scores for each simulation with respect to IMERG. On the x-axis the window size used for the FSS computation.

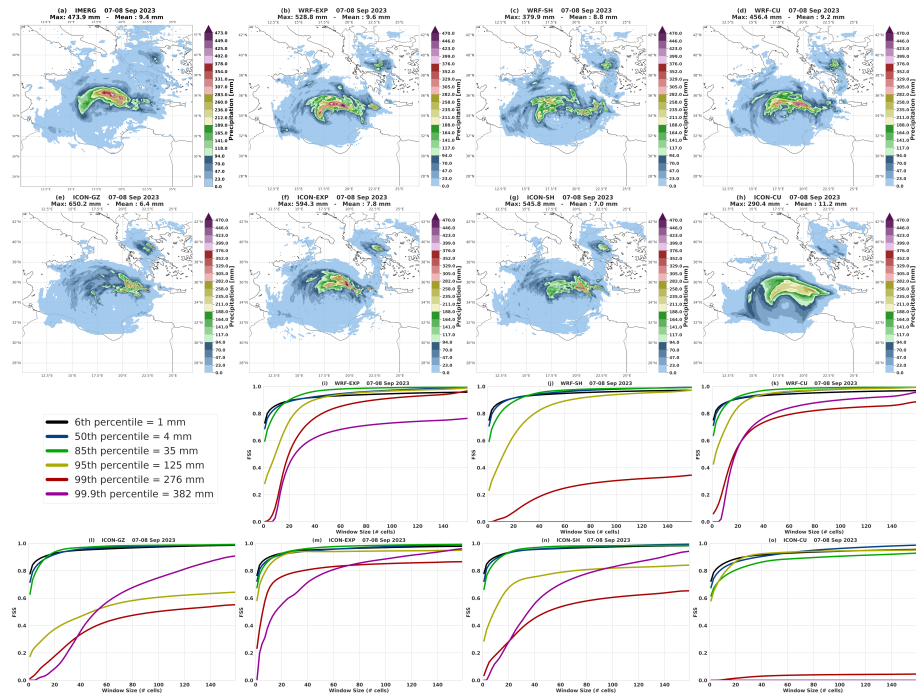


Figure A2. Accumulated total precipitations during transition phase (from 2023-09-07 to 2023-09-08). (a) Observations from IMERG, (b) WRF-EXP, (c) WRF-SH, (d) WRF-CU, (e) ICON-GZ, (f) ICON-EXP, (g) ICON-SH, (h) ICON-CU. Color shades as for FSS percentile thresholds. FSS scores for each simulation with respect to IMERG. On the x-axis the window size used for the FSS computation.

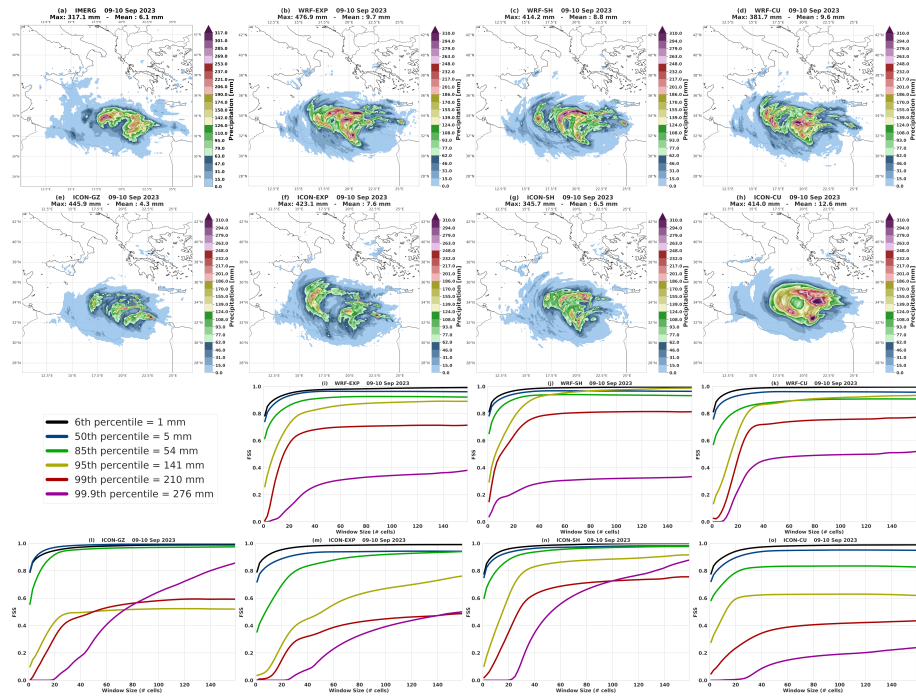


Figure A3. Accumulated total precipitations during tropical phase (from 2023-09-09 to 2023-09-10). (a) Observations from IMERG, (b) WRF-EXP, (c) WRF-SH, (d) WRF-CU, (e) ICON-GZ, (f) ICON-EXP, (g) ICON-SH, (h) ICON-CU. Color shades as for FSS percentile thresholds. FSS scores for each simulation with respect to IMERG. On the x-axis the window size used for the FSS computation.

Model-minus-IMERG difference maps of the event-total accumulated precipitation for all seven configurations. Both model fields and IMERG were first conservatively regridded to the common 0.1° verification grid, the same used for the FSS, so that the differences are not contaminated by resolution mismatch. The maps show that the dominant error structure along the cyclone track is a dipole pattern centred on the observed rainfall maxima, indicating that displacement, rather than a systematic intensity bias, drives most of the local disagreement during the marine phases. Over Greece, the WRF configurations overestimate accumulation along the windward slopes of the Pindus range, while the ICON configurations underestimate precipitation on the southern flank of the cyclone during the transition phase. The dipole amplitudes are smallest in the SH configurations of both models, consistent with sub-grid transport distributing convective precipitation more evenly and limiting grid-scale moisture convergence.

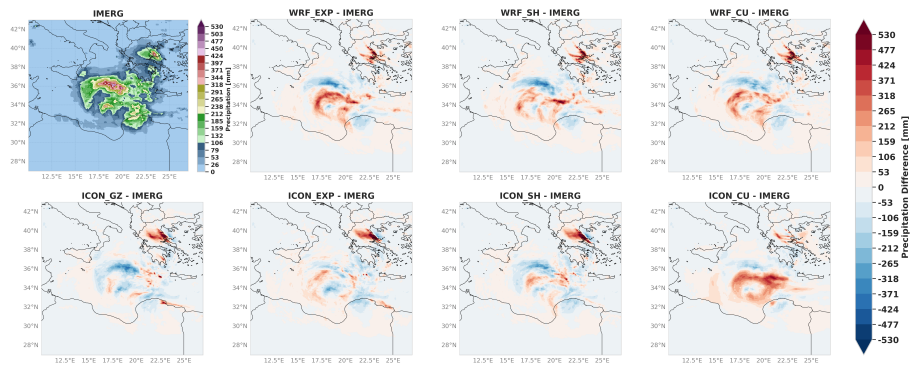


Figure A4. Precipitation field difference Model-IMERG

We computed the frequency distributions of event-total precipitation for all simulations and for IMERG. To avoid areal-averaging artefacts, all distributions are evaluated on the common 0.1° verification grid. The probability density function (PDF) is computed for the bins $[1, 10, 50, 100, 200, 300, \dots, 1800]$. All configurations reproduce the observed distribution up to approximately the 50th percentile, with a consistent behaviour up to 85th percentile. Above it, the model tails are heavier than the IMERG tail; the excess is largest for ICON-EXP and smallest for the ICON-CU configuration, in agreement with the FSS results at the 99th and 99.9th percentile thresholds. Part of the tail disagreement lies in the reference itself: IMERG Final applies a monthly climatological gauge adjustment and underestimated the observed maxima over Thessaly during this event (Flaounas et al., 2025), so the model tail excess over land should be read as an upper bound. Overall these results give no more informations than what already presented in the previous analysis.

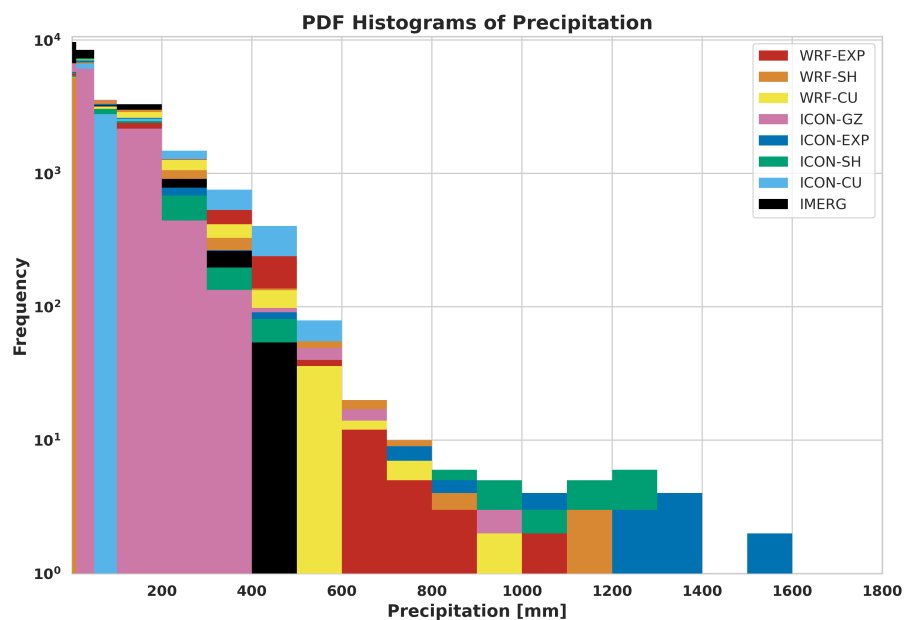


Figure A5. Precipitation PDF

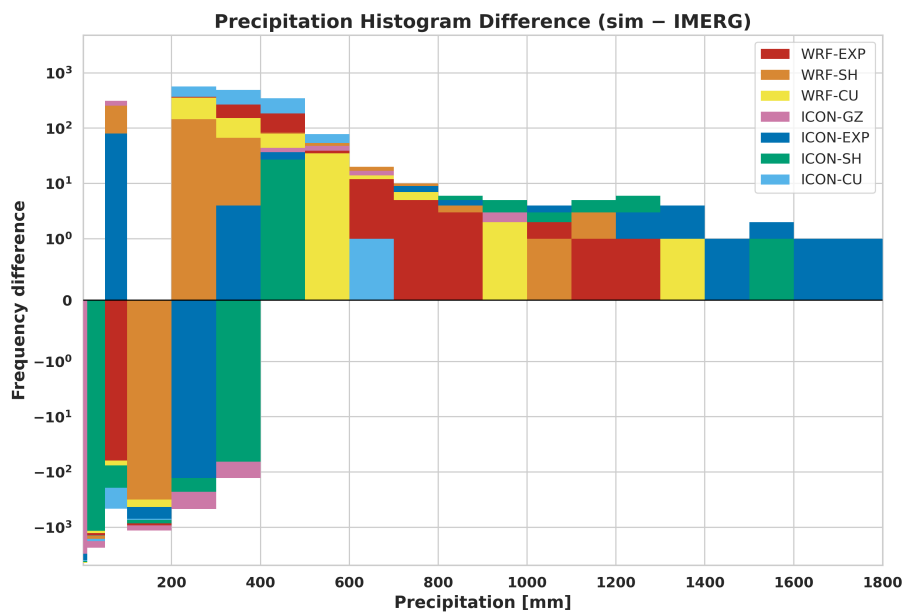


Figure A6. Precipitation PDF Model-IMERG

A2 TASM axisymmetry and displacement

To assess the degree of asymmetry averaged over the annular domains, we calculated the azimuthal standard deviation of the equivalent potential temperature (θ_e) for each radius and vertical level.

The target control (CTRL) simulation is the WRF–SH one, as it provides the best overall result. The azimuthal mean is defined as:

$$\overline{\theta_e}(r, z) = \frac{1}{N} \sum_{i=1}^N \theta_e(r, \phi_i, z)$$

795 Consequently, the azimuthal dispersion is computed as:

$$\sigma_\phi(r, z) = \sqrt{\frac{1}{N} \sum_{i=1}^N [\theta_e(r, \phi_i, z) - \overline{\theta_e}(r, z)]^2} \quad (\text{A1})$$

where r is the radial distance from the cyclone center, z is the vertical level, ϕ_i represents the i -th azimuthal point within the annulus, and N is the total number of grid points within the annular ring. The parameter $\sigma_\phi(r, z)$ provides a quantitative, localized estimate of the field's asymmetry.

800 The axisymmetry assumption holds remarkably well in the inner core of the cyclone. In this central region, σ_ϕ generally remains below 2 K Fig. A7. As expected, the asymmetry naturally increases moving radially outward. Specifically, we identify a localized maximum of asymmetry at approximately 180 km from the center at the 500 hPa level. Beyond a radius of 250 km, the σ_ϕ values exceed 3, indicating a rapid degradation of the symmetric structure. Therefore, we can confidently state that the axisymmetry assumption underlying the TASM is physically robust and well-justified within the inner core of the medicane
805 (up to a radius of 200–250 km).

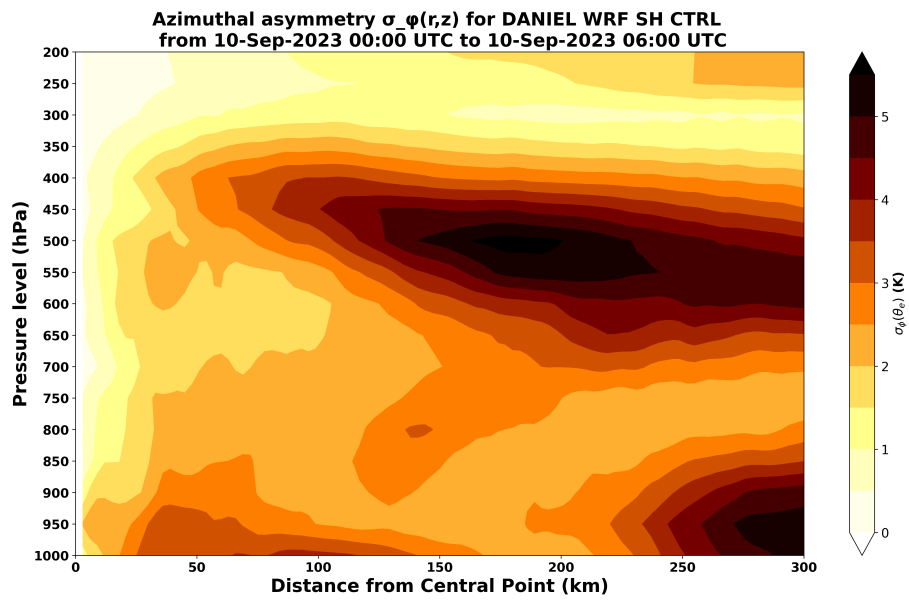


Figure A7. Azimuthal asymmetry (Eq. A1) in the CTRL configuration (WRF-SH). Warm color shades from white to dark brown proportionally to the asymmetry.

To address the propagation of the 15 km positional uncertainty into the annular means, we performed a dedicated spatial sensitivity test on the CTRL simulation dataset. The tracked center of the cyclone was artificially displaced by 0.14° (approximately 15 km) in the four cardinal directions (North, South, East, and West). For each of these displaced configurations, we recalculated the TASM and compared it against the original, centered CTRL output Fig.A8.

810 The statistical comparison reveals that the 2D spatial correlation between the displaced TASM fields and the original reference field is exceptionally high, ranging strictly between 0.9990 and 0.9992.

Therefore, we can state that a center-position uncertainty of 15 km does not significantly propagate into the annular means. The azimuthal averaging effectively dampens the localized spatial noise caused by minor misplacements. Consequently, the TASM calculation proves to be highly resilient to typical tracking algorithm uncertainties, guaranteeing that the resulting thermal

815 anomaly profiles are physically reliable and not artifacts of tracking precision.

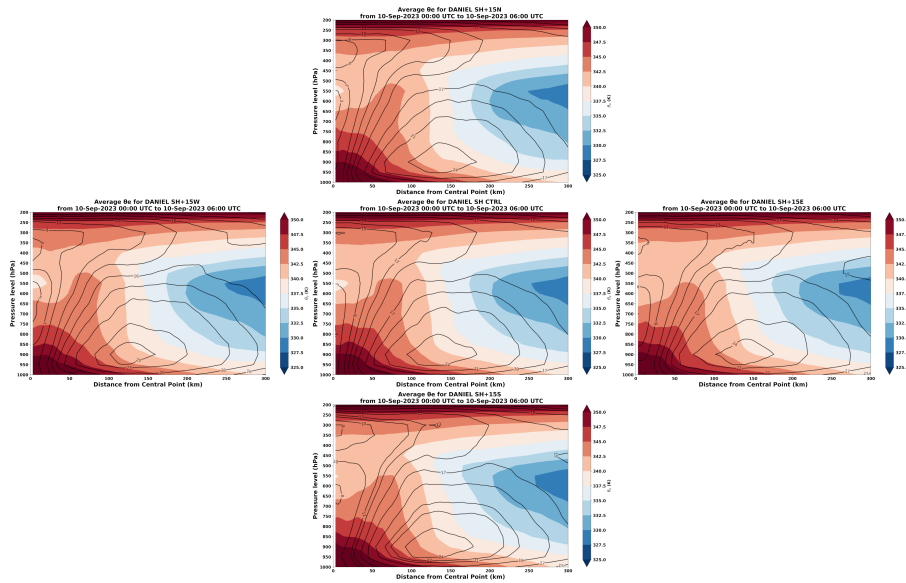


Figure A8. Displacement test in the computation of TASM. CTRL run in the center and cardinal displacement respectively Northward (up), Southward (below), Eastward (right) and Westward (left). Same representation of Fig.10