Trends in Western North Pacific Tropical Cyclone Intensity Change Before Landfall

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This study investigates the long-term trend in the average 24-h intensity change (ΔV₂₄) of western North Pacific (WNP) tropical cyclones (TCs) before landfall during June-November for the period from 1970–2019. We find a significant increasing trend in basin-averaged ΔV₂₄ during 1970–2019. The increase in ΔV₂₄ is significant over the northern South China Sea (17.5°-25°N, 107.5°-120°E) and to the east of the Philippines (7.5°-15°N, 122.5°-132.5°E), implying a slower weakening rate before landfall for the South China Sea and an increased intensification rate before landfall for the region east of the Philippines. We find a significant linkage between changes in ΔV₂₄ and several large-scale environmental conditions. The increased ΔV₂₄ before landfall in the above two regions is induced by a warmer ocean (e.g., higher sea surface temperatures, maximum potential intensity and TC heat potential) and greater upper-level divergence, with a moister mid-level atmosphere also aiding the ΔV₂₄ increase east of the Philippines. Our study highlights an increasing tendency of ΔV₂₄ before landfall, consistent with trends in ΔV₂₄ over water and over land as found in previous publications.

Keywords: tropical cyclone, intensity change, western North Pacific, before landfall, environmental changes

INTRODUCTION

Tropical cyclones (TCs) are one of the most devastating global natural disasters, inducing large economic losses as well as fatalities for various coastal regions. Among TC metrics, TC intensity change has long been regarded as a major challenge for both the scientific research and operational forecasting communities (Courtney et al., 2019; Hendricks et al., 2019). TC intensity change is the result of a complex interaction between various internal influences that are related to the structure and internal processes of the TC itself and various external influences that are controlled by the large-scale atmospheric and oceanic environment (Hendricks et al., 2019).

Given active research on the relationship between TCs and climate change, there has been an increasing focus on temporal variations in TC intensity change. Bhatia et al. (2019) reported an increasing trend in the mean TC 24-h intensity change (ΔV₂₄) over the globe and for the Atlantic basin specifically since the 1980s. They also found a broadening distribution of ΔV₂₄, due to increasing intensification and weakening rates. Similar changes in the ΔV₂₄ distribution from 1982 to 2019 were shown over the western North Pacific (WNP) in Song et al. (2020). The increasing intensification rate is associated with an increasing proportion of rapid intensification (RI) cases that likely has an anthropogenic warming component (Bhatia et al., 2019). By comparison, there is a linkage between the increasing weakening rate and the increasing proportion of rapid weakening
cases, possibly resulting from increasing sea surface temperature (SST) gradients in the subtropics (Song et al., 2020). Note that all of the above findings are only based on TC records over the open ocean.

By analyzing TC samples over land, Liu et al. (2020) demonstrated a slight decreasing trend in the TC weakening rate after landfall in mainland China during 1980–2018, implying an increasing trend in overland $\Delta V_{24}$. This was attributed to decreasing vertical wind shear (VWS), increasing upper-level divergence and increasing mid-level upward motion (Liu et al., 2020). Over mainland China, the decreasing TC weakening rate is consistent with the increasing decaying timescale of landfalling TCs, as shown in Song et al. (2021). They found a slower decay in the first 24 h after landfall, primarily driven by increasing low-level vorticity in coastal regions of China (Song et al., 2021). Additionally, Li and Chakraborty (2020) reported a slowing trend in the decay of landfalling TCs over the North Atlantic, mainly from a contemporaneous increase in SST.

There are still 24-h TC tracks excluded from consideration in the aforementioned previous publications, which examined either five 6-hourly records occurring over water or five 6-hourly records over land. The samples that have been excluded in previous publications move from water to land during the 24-h period, and are consequently expected to have complex $\Delta V_{24}$s, due to multiple environmental and land surface changes that occur during the 24-h period. The most important of these intensity change questions is the $\Delta V_{24}$ before landfall, which is critical for reducing damage and loss of life. Rappaport et al. (2010) studied the intensity change of TCs making landfall along the U.S. Gulf Coast, in which TC tracks in the 48 h prior to landfall were considered. They found that, on average, category 1–2 (category 3–5) hurricanes strengthened (weakened) before landfall, and this observed trend could be partially explained by environmental conditions.

Up until now, it is still unclear what environmental conditions are related to $\Delta V_{24}$ before landfall over the WNP and their potential long-term trends. The reminder of this study is arranged as follow. Data introduces the data used in this study. Trends in $\Delta V_{24}$ Before Landfall examines the long-term trends in the average $\Delta V_{24}$ before landfall and its contributors. Changes in Environmental Conditions highlights changes in environmental variables and their links to changes in $\Delta V_{24}$ before landfall. A summary is given in Summary.

DATA

WNP TC best track data used in this study are given by the Joint Typhoon Warning Center (JTWC), the Japan Meteorological Agency (JMA), the China Meteorological Administration (CMA) and the Hong Kong Observatory (HKO) including 6-hourly TC central positions and maximum sustained winds, as compiled in the International Best Track Archive for Climate Stewardship (IBTrACS) v04r00 (Knapp et al., 2010). Owing to the relatively low quality of the TC intensity estimates in the best track data prior to the 1970s (Camargo and Sobel, 2005), we focus on the period from 1970–2019. To reduce the uncertainty in detecting weak TCs (e.g., tropical depressions) that are induced by changing observational platforms (Klotzbach and Landsea, 2015), we only consider TCs with a lifetime maximum intensity of at least 34 kt. TCs forming during June-November (JJASON) are analyzed here, accounting for ~85% of the annual total number of WNP TCs (Song and Klotzbach, 2019). Similar to Rappaport et al. (2010) and Zhu et al. (2021), a 24-h track before landfall is identified in this study as when the last record is over land and the previous four 6-h records are all over water. Any 24-h tracks with records labeled as extratropical cyclones in the best track data are removed, in order to minimize the influence of extratropical transition on intensity change. In total, there are 4307 identified 24-h tracks before landfall over the WNP (Figure 1A). The mid-points of these tracks are further...
interpolated onto a 2.5° × 2.5° grid. Our results are not significantly changed if a 5° × 5° resolution is used instead (figure not shown).

Monthly large-scale environmental conditions are provided by the fifth generation European Centre for Medium-Range Weather Forecasts (ECMWF) reanalysis of the global climate (ERA5; Hersbach and Bell, 2020), including SST, 200-hPa temperature, 700–500-hPa relative humidity, 850-hPa relative vorticity, 200-hPa divergence and 850–200-hPa VWS. The original ERA5 data over a grid of 0.25° × 0.25° are re-gridded to a resolution of 2.5° × 2.5°, in order to highlight large-scale features. Maximum potential intensity (MPI; Emanuel, 1988) is calculated from monthly ERA5 data. TC heat potential (TCHP), which measures ocean heat content that is warmer than 26°C (DeMaria et al., 2005), is estimated using monthly subsurface temperature profiles from the ECMWF Ocean Reanalysis System 5 (ORAS5; Zuo et al., 2019) with a resolution of 1° × 1°.

TRENDS IN ΔV₂₄ BEFORE LANDFALL

Figure 1A shows all of the 24-h tracks before landfall over the WNP during JJASON between 1970 and 2019. These tracks are located near the coasts of most East Asian and Southeast Asian countries. While it is typically viewed that TCs weaken as they approach land due to interactions with topography, the average ΔV₂₄ before landfall exhibits obvious spatial inhomogeneities (Figure 1B). There are positive average ΔV₂₄ near the Philippines and to the southeast of Vietnam, indicating that TCs, on average, intensify within 24 h prior to landfall in these regions. While Brand and Blelloch (1973) only examined a limited number of TCs, they reported an average TC intensity increase prior to hitting the Philippines. This TC intensity increase is likely caused by environmental conditions near the Philippines being more characteristic of an oceanic environment. Furthermore, negative average ΔV₂₄ are observed along other WNP coastlines, and the magnitude of the negative ΔV₂₄ generally increases with latitude. This result implies that TCs tend to weaken at a greater rate prior to landfall at higher latitudes, possibly as a result of both lower SSTs and increased VWS at higher latitudes.

Figure 2A displays a significant increasing trend for the JJASON average ΔV₂₄ during 1970–2019, with increasing trends of 0.11 kt yr⁻¹ (p < 0.01), 0.11 kt yr⁻¹ (p = 0.04), 0.09 kt yr⁻¹ (p < 0.01) and 0.06 kt yr⁻¹ (p = 0.04) for the best track data from the JTWC, the JMA, the CMA and the HKO, respectively. Given that these increasing trends are relatively consistent between the four agencies, we use the JTWC dataset for all of the remaining analysis.

To examine the relative contributions of incorporated variables to the overall ΔV₂₄ change, a decomposition of the average ΔV₂₄ in each year (ΔV₂₄̄) is performed as:

\[ \Delta V_{24} = \Delta V_{24,\text{Total}} - \Delta V_{24,\text{Intensity}} - \Delta V_{24,\text{Frequency}} - \Delta V_{24,\text{Nonlinear}} \]

\[ \Delta V_{24,\text{Total}} = \Delta V_{24,\text{Term I}} + \Delta V_{24,\text{Term II}} + \Delta V_{24,\text{Term III}} \]

where ΔV₂₄̄ is the total change in intensity, ΔV₂₄,Term I is the intensity effect, ΔV₂₄,Term II is the frequency effect, and ΔV₂₄,Term III is the nonlinear effect.
Changes in ΔV_{24} \text{ (unit: kt)} before landfall during JJASON from 1970–1994 to 1995–2019. Black crosses refer to differences significant at the 0.05 level based on a Student’s t-test. Green boxes highlight Regions A and B.

FIGURE 3

\[ \Delta V_{24m}(t) = \sum \lambda \phi p(\lambda, \phi, t) \times \Delta V_{24}(\lambda, \phi, t). \]  

Here, \( \lambda, \phi \) and \( t \) refer to latitude, longitude and year, respectively. \( p(\lambda, \phi, t) \) denotes the spatial distribution of TC occurrence over a 2.5° × 2.5° grid, while \( \Delta V_{24}(\lambda, \phi, t) \) represents the average \( \Delta V_{24} \) in the corresponding grid. Eq. 1 can be further written as:

\[ \Delta V_{24m}(t) = \sum \lambda \phi p(\lambda, \phi, t) + p^c(\lambda, \phi, t) \times [\Delta V_{24}(\lambda, \phi) + \Delta V_{24}^{\text{clim}}(\lambda, \phi, t)]. \]  

The superscripts “c” and “a” refer to the climatological average value and the anomaly relative to the climatology, respectively. Finally, Eq. 2 is decomposed as:

\[ \Delta V_{24m}(t) = \sum \lambda \phi p^c(\lambda, \phi, t) \times \Delta V_{24}^{\text{clim}}(\lambda, \phi) \]

\[ + \sum \lambda \phi p^a(\lambda, \phi, t) \times \Delta V_{24}^{\text{clim}}(\lambda, \phi, t) \]

\[ + \sum \lambda \phi p^a(\lambda, \phi, t) \times \Delta V_{24}(\lambda, \phi, t). \]  

In Eq. 3, because the climatology term does not vary with time, the temporal change in the average \( \Delta V_{24} \) can only be influenced by the three other terms, namely the frequency effect, the intensity effect and the nonlinear effect.

There is a significant increasing trend in \( \Delta V_{24} \) related to the intensity effect, with a slope of 0.06 kt yr\(^{-1}\) (\( p < 0.01 \)), accounting for approximately one-half of the total \( \Delta V_{24} \) trend (Figure 2B). By contrast, we find no significant trend in \( \Delta V_{24} \) related to the frequency effect, whose rate is lower than the total \( \Delta V_{24} \) trend by one order of magnitude (Figure 2C). The trend in \( \Delta V_{24} \) related to the nonlinear effect is not significant, although its magnitude is comparable to that related to the intensity effect (Figure 2D). The reason that the nonlinear trend is not significant may be due to the larger standard deviation of this term (3.5 kt) relative to the intensity effect (1.7 kt). These results indicate that the intensity effect is the primary driver of the long-term changes in the total \( \Delta V_{24} \) while the frequency effect and the nonlinear effect have a lesser impact.

Figure 3 displays the differences in \( \Delta V_{24} \) before landfall during JJASON between two sub-periods (1970–1994 and 1995–2019). Similar features are obtained if long-term \( \Delta V_{24} \) trends from 1970–2019 are displayed instead (figure not shown). Significant increases in \( \Delta V_{24} \) are concentrated over two regions: one is located over the northern South China Sea (SCS) (Region A: 17.5°–25°N, 107.5°–120°E), while the other is located to the east of the Philippines (Region B: 7.5°–15°N, 122.5°–132.5°E). Given the climatological \( \Delta V_{24} \) distribution in Figure 1A, the \( \Delta V_{24} \) increase in Region A (Region B) implies a slower weakening (stronger intensification) of TCs before landfall. By comparison, changes in \( \Delta V_{24} \) over other regions are of a lower magnitude and are less significant. We thus conclude that the increase in basin-averaged \( \Delta V_{24} \) is primarily induced by the \( \Delta V_{24} \) increases over Regions A and B.

CHANGES IN ENVIRONMENTAL CONDITIONS

Although TCs make landfall near the end of the identified 24-h tracks, they are over the ocean during most of the 24-h period. Consequently, these 24-h tracks are more likely influenced by the environment over water than over land. Figures 4A–E illustrates changes in thermodynamic conditions during JJASON from 1970–1994 to 1995–2019. There are significant increases in SST, MPI and TCHP over almost all of the WNP (Figures 4A–C), consistent with the global warming that has occurred since the middle of the last century. Compared with the period from 1970–1994, higher SST, MPI and TCHP in 1995–2019 inhibit the decaying of TCs before landfall over Region A and favor the intensification of TCs before landfall over Region B. There are no significant changes in 200-hPa temperature over Regions A and B from 1970–1994 to 1995–2019 (Figure 4D). While 200-hPa temperature has also increased, the increases in SST and 200-hPa temperature are of comparable magnitude, yielding a thermodynamic environment that is more favorable for TC intensification (Tuleya et al., 2016). In general, the mid-level atmosphere has become moister over the ocean and drier over land from 1995–2019 relative to 1970–1994 (Figure 4E). Although the 700–500-hPa relative humidity has only changed slightly over Region A, relative humidity has increased significantly over Region B. A moister environment is favorable for TC development and intensification, helping to increase the TC intensification rate before landfall over Region B. Given that MPI is a function of SST and the profiles of atmospheric temperature and humidity (Emanuel, 1988), the increasing MPI over Region A is primarily induced by increasing SST, while the increasing MPI over Region B is jointly driven by increasing SST and the moistening atmosphere.
FIGURE 4 | Changes from 1970–1994 to 1995–2019 for JJASON averages of the following environmental variables: (A) SST, (B) TCHP, (C) MPI, (D) 200-hPa temperature, (E) 700–500-hPa relative humidity, (F) 850-hPa relative vorticity, (G) 200-hPa divergence, and (H) 850–200-hPa VWS. Black crosses denote values significant at the 0.05 level based on a Student’s t-test. Green boxes denote the two focus regions of this study.
Figures 4F–H displays differences in dynamic variables during JJASON between 1970–1994 and 1995–2019. There are no significant changes in 850-hPa relative vorticity and 850–200-hPa VWS over Regions A and B (Figures 4F,H), indicating that changes in these variables related to the low-level environmental circulation have only had minor impacts on TC intensity changes before landfall. By contrast, significant increases in 200-hPa divergence are found over both Regions A and B (Figure 4G). The enhanced upper-level divergence favors TC development and subsequently increases ΔV24 before landfall. The above environmental variables exhibit similar features if their long-term trends from 1970–2019 are displayed (figure not shown).

To confirm the relationship between ΔV24 before landfall and environmental variables over Regions A and B, Figure 5 examines JJASON correlations between ΔV24 and environmental variables from 1970 to 2019. Over Region A, there is a significant increasing trend in the average ΔV24, with a slope of 0.18 kt yr−1 (p < 0.01). This increasing trend is much larger than the trend in the basinwide average ΔV24. Changes in average ΔV24 significantly correlate with changes in SST and 200-hPa divergence, with correlation coefficients of 0.34 (p = 0.02) and 0.38 (p < 0.01), respectively (Figure 5A). However, there is no significant correlation between the changes in average ΔV24 and 700–500-hPa relative humidity (r = −0.21; p = 0.14).

By comparison, over Region B, the average ΔV24 shows a significant increasing trend of 0.17 kt yr−1 (p = 0.04). The change in average ΔV24 is significantly correlated with changes in SST, 700–500-hPa relative humidity and 200-hPa divergence, with correlation coefficients of 0.35 (p = 0.01), 0.31 (p = 0.03) and 0.30 (p = 0.03), respectively (Figure 5A). Additionally, although SST, MPI, TCHP and 200-hPa divergence significantly increase to the south of Japan (27.5°–35°N, 127.5°–135°E), there are no significant changes in the average ΔV24 before landfall. The lack of trend may be related to the low number of TCs occurring over this region (1.2 TCs per year on average). This low TC frequency can lead to large variability in the annual change of average ΔV24, subsequently reducing the significance of the long-term trend.

SUMMARY

This study investigates long-term trends in average ΔV24 before landfall during June-November from 1970–2019. After
identifying 4307 24-h tracks before landfall, we display the climatological spatial distribution of their $\Delta V_{24}$. On average, TCs intensify before landfall near the Philippines and to the southeast of Vietnam, while they weaken before landfall over other coastal regions. There is a significant increasing trend in basin-averaged $\Delta V_{24}$ during 1970–2019, regardless of best track dataset used to identify TCs. This increasing trend is primarily caused by changes in $\Delta V_{24}$ over individual grids, while it is only weakly influenced by changes in the TC occurrence distribution. We find that $\Delta V_{24}$ before landfall increases significantly over the northern SCS (Region A: 17.5°–25°N, 107.5°–120°E) and to the east of the Philippines (Region B: 7.5°–15°N, 122.5°–132.5°E). This implies a weakening decay rate over Region A and an increased intensification rate over Region B for WNP TCs before landfall.

The changes in $\Delta V_{24}$ before landfall over Regions A and B correlate well with changes in several large-scale environmental variables. The greater $\Delta V_{24}$ before landfall over Regions A and B can be linked to a warmer ocean (e.g., higher SST, MPI and TCHP) and greater upper-level divergence in 1995–2019 than in 1970–1994. By comparison, the greater $\Delta V_{24}$ before landfall over Region B is likely also a result of a moister mid-level atmosphere. Our study highlights an increasing tendency of $\Delta V_{24}$ in the WNP before landfall, consistent with trends in $\Delta V_{24}$ over water and over land (Bhatia et al., 2019; Liu et al., 2020).

**DATA AVAILABILITY STATEMENT**

The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author.

**AUTHOR CONTRIBUTIONS**

QL, JS, and PK contributed to conception and design of the study. QL and JS organized the database and performed the statistical analysis. QL wrote the first draft of the article. JS and PK revised the article. All authors contributed to manuscript and approved the submitted version.

**FUNDING**

This work was jointly funded by the National Key Research and Development Program of China (2018YFC1507305), the National Natural Science Foundation of China (61827901 and 42175007) and the China Postdoctoral Science Foundation (2020M680789). Klotzbach would like to acknowledge financial support from the G. Unger Vetlesen Foundation.

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