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Usability of best track data in climate statistics in the western North Pacific.

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1

2 Abstract

3 Tropical cyclone (TC) activity for the last three decades shows strong discrepancies, deduced
4 from different best track data (BTD) sets for the western North Pacific (WNP).

5 This study analyses the reliability of BTD sets in deriving climate statistics for the WNP.
6 Therefore TC lifetime, operational parameters (CI-number) and tracks are compared (for TCs
7 identified concurrently) in BTD provided by the Joint Typhoon Warning Center (JTWC),
8 Japan Meteorological Agency (JMA), and the China Meteorological Administration (CMA).

9 The differences between the BTD are caused by varying algorithms used in weather
10 services to estimate TC intensity. Available methods for minimizing these discrepancies are
11 not sufficient. Only if intensity categories 2-5 are considered as a whole, do trends for
12 annually accumulated TC-days show a similar behaviour.

13 The reasons for remaining discrepancies point to extensive and not regular usage of
14 supplementary sources in JTWC. These are added to improve the accuracy of TC intensity
15 and center position estimates. Track- and CI- differences among BTD sets coincide with a
16 strong increase in numbers of intense TC-days in JTWC. These differences are very strong in
17 the period of intensive improvement of spatial-temporal satellite coverage (1987-1999).

18 Scatterometer-based data used as a reference show that for the tropical storm phase
19 JMA provides more reliable TC intensities than JTWC. Comparisons with aircraft
20 observations indicate that not only homogeneity but also a harmonization and refinement of
21 operational rules controlling intensity estimations should be implemented in all agencies
22 providing BTD.

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1 Keywords: tropical cyclone, best track data, climate change, Northwest Pacific

1.2 Introduction

3 In recent years, tropical cyclone (TC) activity, which poses a risk for coastal
4 populations, gained much attention in the environmental research community (Emanuel,
5 2005; Landsea, 2005; Webster et al., 2005; Wu, 2006). The long-term variability in TC
6 activity became a subject of interest in atmospheric science pointing to changes in
7 atmospheric rotational flow, vertical wind shear or sea surface temperature (SST) over the last
8 decades (Chan et al., 2004; Trenberth, 2005). Using tropical cyclone “best track data”
9 (hereafter referred to as BTD) sets Webster et al. (2005) and Emanuel (2005) claimed there
10 would be an increase in the occurrence of the most intense TCs in the western North Pacific
11 (WNP). However, according to Wu et al. (2006), who used several BTD sets provided by
12 different institutes, neither the numbers of the most intense TCs nor the power dissipation
13 index (PDI) defined by Emanuel (2005) show an increasing tendency.

14 Comparing three BTD sets, Ren et al. (2011) confirmed increasing TC tendencies for
15 the Joint Typhoon Warning Center (JTWC) BTD, but they found decreasing tendencies in the
16 data of the Japan Meteorological Agency (JMA) and the China Meteorological
17 Administration (CMA). Kamahori et al. (2006) found increasing numbers of TC-days for
18 categories 2 to 3 of the Saffir–Simpson Hurricane Scale (hereafter SSHS) and decreasing
19 numbers in higher categories for JMA, while opposite trends were detected for the JTWC data
20 set. All these studies indicate a great dependency of the detected TC trends on the chosen
21 BTD, pointing to data inhomogeneity and quality deficiencies in the WNP region.

22 Knaff and Sampson (2006) considered any detected intensity trend questionable
23 before reanalyses employing data sets of TC intensity estimated with alternative techniques
24 are incorporated. Others attempted to identify the reasons for the differences between BTD
25 which affect TC activity trends (Kamahori et al., 2006; Nakazawa and Hoshino, 2009; Song et

al., 2010). Many studies highlighted the different operational procedures used by the individual meteorological agencies to estimate TC intensity as a main cause for differing TC activity results. Knapp and Kruk, (2010) attempted to minimize discrepancies among BTD by applying unified algorithms to operational data from all centers, resulting in more comparable BTD sets.

In this paper we assess the reliability of BTD in deriving TC activity trends. In the first part of the study we evaluate the skill of current solutions for achieving homogeneity between the individual data sets. In the second part, the remaining discrepancies between BTD sets are analysed and evaluated using independent reference data sets. All data and methods used are described in section 2. Results and discussion of comparisons are presented in section 3. Section 4 summarizes and concludes the article.

Data and methods

Four different BTD sets were analysed in this study. They were provided by the following independent agencies: the China Meteorological Administration (CMA, www.typhoon.gov.cn), the Regional Specialized Meteorological Center (RSMC), Tokyo of the Japan Meteorological Agency (JMA, www.jma.go.jp/jma/jma-eng/jma-center/rsmc-hp-pub-eg/besttrack.html) and the Joint Typhoon Warning Center (JTWC, www.usno.navy.mil/NOOC/nmfc-ph/RSS/jtwc/best_tracks/wpindex.html). In addition the International Best Track Archive for Climate Stewardship, (IBTrACS, <http://www.ncdc.noaa.gov/oa/ibtracs/index.php?name=ibtracs-data>) was used. This product combines BTD from different operational centers to create a global best track dataset. Although IBTrACS can not serve as independent data, it provides useful information as it gives a merged BTD solution for which a data quality control was applied. BTD sets for the WNP contain TC centre, maximum sustained wind and central pressure at 6-hour intervals.

JTWC and CMA intensity values start with Tropical Depression (TD) strength, and JMA starts with Tropical Storm (TS) category.

From 1977 JMA began recording maximum sustained wind speeds using the Dvorak technique (hereafter referred to as DT) (Dvorak, 1972, 1973, 1975). Since 1987, when aircraft reconnaissance flights ended in the WNP, this method became the main tool for compiling BTD sets. The technique estimates TC position and intensity using visible and infrared imageries from geostationary and polar-orbiting weather satellites. However, procedural rules to process the data for BTD within meteorological agencies were evolving differently. Dvorak parameters (T-number and Current Intensity (CI) number), estimated operationally on basis of identified cloud patterns, are related to TC intensity through conversions which were independently established for differing wind speed definitions in each operational center. While the JTWC uses 1-minute mean sustained 10 m wind speed, as designed originally by the Dvorak technique, other agencies use 10-min averaged values. JMA established a new conversion table in 1990 (Koba et al., 1991) which transfers operational parameters (CI) directly to TC intensity described as 10-min maximum sustained wind speed.

The CMA data set specifies intensity in terms of “2-min mean maximum sustained wind speed (m s^{-1}) near the storm centre”. However, this procedure contradicts the description in Yu et al. (2007) which states that the CMA agency uses an empirically established linear relationship between 1-min and 10-min averaged values and multiplies wind values by a factor of 0.871. The assumed application of a 10-min-average definition in the CMA data set is supported by findings of relatively small differences among JMA and CMA (Knapp and Kruk (2010)). IBTrACS data use 10-min sustained wind speed.

In order to evaluate the BTD additional observational data sets were tested for their ability to serve as a reference. Blended Sea Winds provided by the National Oceanic and Atmospheric Administration's National Climatic Data Center (<http://www.ncdc.noaa.gov/oa/rsad/air-sea/seawinds.html>), denoted as “NOAA”) contain

1 ocean surface wind speed on a global 0.25° grid in 6 hourly time steps. The data are created
2 by blending observations from multiple satellites with a simple spatial-temporally weighted
3 interpolation. The quality of the blended product is related to the accuracy of the input data
4 and sampling scheme of the observations. The number of long-term US satellites providing
5 wind observations increased from one in 1987 to five in 2000. In this study years 2000 to
6 2008 were analysed as they constitute a rather homogeneous temporal and spatial coverage.
7 For this period wind observations are retrieved from: Quik Scatterometer (QuikSCAT),
8 SSM/I (DMSP Special Sensor Microwave/Imager), AMSR-E (Advanced Microwave
9 Scanning Radiometer of NASA's Earth Observing System) and the Tropical Rainfall
10 Measuring Mission (TRMM) Microwave Imager (TMI). Scatterometers measure
11 instantaneous ocean surface wind vectors at 10m height with a grid-typical resolution of 25
12 km and are widely used in operationally prepared analyses and forecasts (Bourassa, 2010;
13 Brennan, 2009; Hoffman and Leidner, 2004). They are intended to provide accurate ocean
14 surface winds in all weather conditions except for rain conditions that occur often during high
15 winds. QuikSCAT data, evaluated against buoys, is adhered to an 8-minute average.
16 QuikSCAT winds were shown (Brennan et al., 2009) to have high skills in intensity
17 estimation for tropical storms strength. However, enhanced backscattering by rain may
18 introduce a positive bias during tropical depressions and rain attenuation causes large negative
19 biases for very high winds. Microwave observations flagged as contaminated by precipitation
20 were excluded from the analysis.

21 As reference data for the TCs of the strongest intensity, aircraft measurements were
22 used. For the analysed period 2000-2008 the THORPEX Pacific Asian Regional Campaign
23 (TPARC-2008) aircraft campaign took place in the WNP, which provided measurements of
24 wind speed during TC events. Observations were obtained from Stepped-Frequency
25 Microwave Radiometer (SFMR). Additionally we used the measurements from a field
26 experiment in 2010: Impacts of Typhoons on the Ocean in the Pacific (ITOP-2010). The

databases for both campaigns are available online:
http://www.aoml.noaa.gov/hrd/data_sub/hurr.html.

2.1 Quantifying TC trend differences derived from BTD sets

TC trends for the period 1977-2008 were derived from several BTD sets and compared in the form of annual number of TC-days categorized by the SSHS scale. The analysis is constrained to TC observations recorded concurrently in all independent BTD sets. This excludes contributions of differing TC frequency among BTD sets to trend discrepancies and enables the identification of the reasons for differences in estimated intensity. Discrepancies among trends derived from 1-min (JTCW) and 10-min (JMA) sustained wind speed are discussed with regard to the impact of intensity definition on the derived climate statistics. The accuracy and effectiveness of two methods unifying wind definitions is assessed with respect to minimizing trend discrepancies.

The methods adjusting TC intensity definitions from 10-to-1 min averaging period were applied to JMA and CMA. The first method is based on the statistical, linear relationship between 10-min and 1-min averaged intensity (Atkinson (1974)). The data from JMA and CMA (for CMA a 10-min average is assumed as stated in the previous section) multiplied by a factor of 1.14 are hereafter referred to as JMA*1.14, and CMA*1.14.

Knapp and Kruk (2010), Song et al. (2010), and Wu et al. (2006) highlighted the problem of different algorithms used among the various BTD to convert CI parameters (derived from satellite imageries) to wind speed. An alternative method was proposed by Knapp and Kruk (2010) and Kruk et al. (2011), reversing intensity values back to operational parameters (CI) and then applying a single conversion table to all data sets resulting in more homogenous intensity values. Following these guidelines, the JMA data set was reverted to CI numbers, using the conversion tables described in Koba et al. (1991). In a second step, we derive wind speed from CI numbers by applying the original Dvorak conversion table

(Dvorak, 1984) used in JTWC (hereafter DT conversion). It is possible that the Koba conversion table was applied only to intensity records starting in 1991 and previous years were not updated to the new procedures (Nakazawa and Hoshino, 2009). However, the remapping method using the Koba conversion table was applied for the complete analysis. Consequently, years before 1987 should be analysed with extreme caution and have only minor impact on the conclusions derived in this article.

The remaining reasons for BTD trend discrepancies are examined by comparing data sets with the same wind speed definition (JTWC and JMA/CMA adjusted to 1-min averaging period). The statistical analysis additionally includes yearly mean differences for TC center locations, annual distributions of differences between BTD sets for CI-numbers, and TC center locations. The difference in TC location is estimated by a measure of distance (ΔP) between two geographical points (x_1, y_1) and (x_2, y_2) on the Earth's surface:

$$\Delta P = r_0 \cdot \cos^{-1} \{ \sin(y_1) \cdot \sin(y_2) + \cos(y_1) \cdot \cos(y_2) \cdot \cos(x_1 - x_2) \}, \quad (1)$$

x and y are longitude and latitude, r_0 is the radius of the Earth.

2.2 BTD-reference data comparison methods

Independent reference data were employed to evaluate the remaining discrepancies between BTD sets. Due to a positive bias which occurs in QuikSCAT data for tropical depressions (Hoffman and Leidner, 2004) and frequently changing procedures in operational centers to identify this phase, the analysis focuses on concurrent records in BTD sets during tropical storm stage. As JTWC and JMA provide information about conversion tables in use, we use the JMA data set remapped to 1-min averaged wind speed using the DT table (as described in the previous section). Concurrent TC observations in BTD were compared with the NOAA wind data for the period 2000-2008, when QuikSCAT had a large impact. To derive maximum TC wind speeds from NOAA, the center positions given by JMA were used. TC circulation in developed systems vanishes at a finite horizontal radius with an upper

boundary of approximately 1000 km (Dean et al., 2009). For small, developing or already dissipated cyclonic systems, it was assumed that the maximum wind speed is within a 500 km radius around a given location. Maximum intensities between the two data sets were compared for all concurrent TC cases.

As microwave signal is vulnerable to heavy rain conditions, the NOAA data exclude such values of reduced accuracy. Therefore time steps with a number of missing values around a TC centre potentially high enough to mask a region of maximum wind speeds were also excluded from the comparison.

For the comparison of the highest intensity typhoons the SFMR observations were used. Observations were obtained during several flights targeting TC centers of typhoons Sinlaku (2008), Jangmi (2008) and Megi (2010). SFMR measures wind speed values in 1 s intervals. To use these wind speeds compatible with BTD, the values were used in two forms: averaged over a 10 second and 1 minute interval. Similar to the previous method, the value of the maximum wind speed was derived by choosing the highest value within a certain radius from the TC center given by JMA.

Results and discussion

3.1 Are the current methods able to minimize discrepancies among TC activity trends in BTD sets?

The damage potential posed by TCs depends on their frequency and duration. To evaluate the skill of the methods in reducing the differences between TC activity trends, we examined the annually integrated TC lifetime for original and modified BTD sets. The analysis was conducted for the period from 1977 to 2008 and only for concurrent observations. Therefore the total TC-day number is the same for every data set and differs

only in the number of records falling into individual intensity categories. The values in the categories of the highest winds are the most significant for socio-economic consideration. Therefore we focused on categories 2 to 5 and TS separately.

Figure 1a presents annually accumulated records of TC-days for categories 2-5. Original data sets are IBTrACS, JMA and JTWC (reporting 1-min sustained wind speed). Data sets adjusted to a 1-min averaging period are JMA*1.14 and CMA*1.14 (which result from applying a multiplication factor to JMA and CMA) and JMADT (where a remapping method – using the original Dvorak (1984) conversion table - was applied to JMA CI-numbers). JMA, IBTrACS and CMA (not shown) show very similar TC-day numbers, with slightly higher numbers for CMA in the first years of the analysis. It was already demonstrated in previous studies (Knapp and Kruk, 2010; Song et al., 2010, F. Ren, 2011) that JTWC and JMA wind speed values show the largest discrepancies among the original data sets. The application of methods to unify the wind averaging period significantly reduced these differences. The average of annual relative differences for the considered period exceeds 0.77 for JMA, and 0.57 for CMA in relation to JTWC. Multiplying JMA and CMA data by 1.14 (JMA*1.14 and CMA*1.14) results in much smaller values, 0.19 and 0.22, respectively. Consequently, JMA*1.14 showed a stronger increasing tendency, similar to JTWC. Recalculating JMA TC intensities from CI parameters with the original Dvorak conversion table (JMADT) also reduced the differences and increased the TC activity trend from 0.18 to 0.45, while JTWC shows the highest trend of 0.65.

To analyse the effectiveness of the methods for different TC intensity categories we analysed the trends for categories 2-3 and 4-5 separately (Figures 1b,c). The JMA TC-days trend for categories 2-3 is high (0.22) and increases to 0.56 when using the multiplicative factor (JMA*1.14). This is twice as high as the JTWC trend. The method has less effect in categories 4-5. Trends in modified data sets (JMA*1.14, CMA*1.14) still retain the decreasing character of 10-min wind speed BTD (JMA, CMA, IBTrACS). In contrast, 1-min

wind speed BTDC (JTCWC) shows upward trends. The results for CMA*1.14 are almost identical to JMA*1.14 which suggests that 10-min-averaged wind speed values were used in CMA (see chapter 2). The results indicate that usage of the multiplicative factor increases intensity values sufficiently to upgrade them to categories 2-3. However it is still too small for upgrading values to categories 4-5, and therefore leads to accumulated TC records in the lower range.

Applying the original Dvorak conversion table to JMA leads to higher numbers of TC-days in both categories 2-3 and 4-5. It increases the trend for categories 2-3 to 0.46, which is already lower than JMA*1.14 (0.56) but still significantly higher than in JTCWC. This method upgrades intensity values to categories 4-5 and reduces partially the differences in TC-day numbers in comparison to JTCWC. While JTCWC features an increasing trend of 0.39 (Figure 1c), JMA*1.14 presents the strongest decrease, JMADT shows no trend.

The interagency differences also change in time for the lower wind categories. Figure 3 presents annually accumulated TC-day records for BTDC sets for the Tropical Storm category, where the highest differences occur for the middle period of the analysis (1987-1998). JMADT shows systematically lower numbers of TC-day records than JMA, as applying the Dvorak conversion degrades over 30% of all records from the TS to the TD category. In contrast, application of the multiplication factor upgrades values to higher categories. Therefore both methods result in smaller TC-day numbers for the TS category. Nevertheless the TC activity tendencies of the analysed records are in good agreement showing a slight increase until the mid-1990s and a decrease for the last decade.

3.2 Impact of unification of conversion tables in BTDC on climate statistics

Knapp and Kruk (2010) found discrepancies among BTDC intensity records to be highly linear, and demonstrated that they can be minimized using a remapping method. Our analysis shows that both methods have the skill to reduce discrepancies among TC activity

1 trends for categories 2-5, and lead to increasing TC trends. Figure 2 presents the functions for
2 converting CI parameters to TC intensity which are used in operational centers in the WNP
3 region. It is visible that for wind speed of category 1 and higher, both conversions – the
4 Dvorak table used in JTWC (DT) and the linear factor ($JMA \times 1.14$) - provide higher wind
5 speed values for the same CI parameter than the Koba conversion. Therefore, the application
6 of such methods reduces the differences in comparison to JTWC due to increasing wind
7 values and due to shifting more low-category TC records towards categories 2-5.

8 When categories 2-3 and 4-5 are regarded separately, the application of the linear
9 relationship has obvious drawbacks, as it introduces high uncertainties to TC trends. The
10 multiplication factor enhances wind speed values linearly, for the whole data set distribution.
11 However, the nonlinear sensitivity of wind speed to the averaging period, which makes
12 Atkinson's (1974) linear relation less accurate, creates the risk of overestimating values in the
13 lower intensity categories (2-3), and underestimating the highest ones. Kamahori et al. (2006)
14 confirmed our findings, showing high discrepancies in trend tendencies between JTWC and
15 linearly modified JMA, but this comparison included all identified TCs in both data sets and
16 not only the concurrent ones. They also found a strong increase in JMA TC-days for
17 categories 2-3, and a decrease for categories 4-5, while JTWC showed opposite tendencies.

18 Applying the original Dvorak conversion considers the non-linear effects of the
19 averaging time interval. The remapping method using the DT conversion reduces the trend
20 discrepancies between JMA and JTWC more efficiently for categories 2-3 and has higher skill
21 in the extreme wind range by upgrading more records to categories 4 and 5. However, the
22 trends in category 4-5 still differ.

23 Song et al. (2010) suggested that the main reasons for differences in BTD intensity
24 over the WNP are different conversion algorithms. Following this hypothesis, applying the
25 same algorithm to all deduced BTD operational parameters should reduce the difference in
26 wind speed to zero, assuming that the same CI parameters were provided by the

1 meteorological agencies. We found that this remapping method leads to enhanced agreement
2 in TC-days for the highest wind speeds, but relatively high differences are still present. This
3 indicates that there are additional contributing factors, which, in the earlier TC intensity
4 estimation stage, cause discrepancies in operational parameters (T, CI).

5 The differences among BTD show temporal variation. High agreement in TC-day
6 records is visible in the first years of the analysis (1977-1987). As a possible explanation,
7 Knapp and Kruk (2010) suggested that the same Dvorak procedures (e.g the same conversion
8 algorithm) were applied for this period. In the second period (1988-1997) numbers and trends
9 among original BTD sets differ a great degree. However, unifying wind speed definitions
10 (application of the Dvorak table to BTD) did not efficiently resolve differences in the highest
11 categories. Discrepancies among BTD sets in this period are increase, very similar to the
12 strong increase of TC-day records in JTWC. In contrast, TC activity for the last decade shows
13 good agreement, and an increasing trend for the categories 4-5 for JTWC, JMA*1.14 and
14 JMADT. We conclude that unifying the conversion algorithms, and thus wind speed
15 definitions, is necessary for an accurate assessment of BTD sets. However, the trend statistics
16 derived from the given datasets remain inconsistent. This requires an explanation of the
17 remaining differences, as offered in the following.

18 3.3. Can the reasons for discrepancies between BTD and the discrepancies 19 themselves be evaluated?

20 Here we focus on the trends derived from JTWC and JMADT in search of additional
21 reasons for the remaining differences. The resulting discrepancies indicate that there are
22 differences among CI numbers provided by the BTD agencies. To visualize the problem,
23 which cannot be resolved by applying the same DT algorithm, two intense typhoons, Isa
24 (1997) and Dianmu (2004), are presented in Figure 4a,b. The figure shows a time series of
25 maximum wind speed given by different BTD. The differences between original 10-min JMA
26 data and 1-min JTWC reach 30 m s^{-1} during peak winds. Adjusting JMA to 1-min wind speed

1 using a multiplication factor reduces the difference to 25 and 20 m s⁻¹ for Isa and Dianmu,
2 respectively. After applying the same Dvorak conversion table a difference of 20 and 15 m s⁻¹
3 still remains, which corresponds to a difference in CI parameters of 1.75 and 1 (Figure 2). For
4 TC Isa, a high discrepancy is noticeable during the whole TC lifetime. For Dianmu, the main
5 differences occur during the highest intensity phase, when the TC in JMADT reaches the 4th
6 category. It is also worthy to note how the multiplication factor shapes the values during the
7 TC lifetime. JMA*1.14 shows higher intensity than JMA/JMADT intensities in the categories
8 TD, TS and 1, but lower intensity than JMADT in the peak categories.

9 Kruk et al. (2011) considered CI parameters for most TCs in the WNP to be almost
10 identical between the BTD agencies, with the 95th percentile varying between 5.75 and 6.25
11 among BTD sets. However, we would like to emphasize that, for the highest intensity
12 categories, noticeable differences are apparent, as shown by two example TCs. Figure 5
13 presents the CI parameter annual differences distribution for JTWC and JMADT,
14 corresponding to the remaining intensity differences, separated into 3 categories: TD-1, 2-3
15 and 4-5. The distribution of TC lifetime discrepancies (Figure 1b,c) reflects the differences of
16 CI numbers. The most pronounced differences are visible for the highest parameters,
17 especially in the second period, 1988-1997. In this period the CI differences were increasing
18 in time and reached the extreme high percentage of CI differences of 2 in 1997. Lower
19 categories, although with smaller CI differences, retained similar features, as did the enhanced
20 TC activity level in the second period, especially in the years 1995-1997. In the early 2000s
21 CI discrepancies are still higher, especially for categories 4-5. Two periods of the strongest CI
22 discrepancies were also identified by Nakazawa and Hoshino (2009), who analysed
23 operational parameters from 1987-2006. They found a significantly higher numbers in JTWC
24 for 1992-1997 and 2000-2005 in comparison to JMA.

25 The reasons for changes in time of CI discrepancies can be related to separately
26 evolving practices and usage of different information sources by operational centers. JMA

1 reports geostationary satellites to be the principal source of TC localization and intensity
2 estimation. In contrast, JTWC emphasize supplementing these data with other: remotely
3 sensed and in – situ observations, that are useful for TC center identification, defining TC
4 structure and providing more direct intensity estimation. The distribution of differences in TC
5 position among BTD shown in Figure 6 might indicate that different satellite - based sources
6 were used for intensity estimation. Figure 6b shows annual means of TC center differences
7 provided by JTWC and JMA. The mean annual differences in TC center position decrease
8 with increasing intensity. The highest discrepancies occur for weak TCs (CI range 1-4.75),
9 where often intensity and centers are difficult to estimate by low- resolved observations. In
10 contrast, there is better agreement in locating the strongest TC centers.

11 The most striking values are visible for the period 1988-1998, when the aircraft
12 reconnaissance era in the WNP was replaced by intensively developing satellite
13 measurements. In that time widely distributed differences in TC locations were up to 150 km
14 with mean annual differences varying between 30-50 km. After 1998 these differences are
15 significantly smaller and do not exceed 30 km.

16 The relationships between BTD trends in these distinct three periods correspond well
17 with those of annual CI differences and TC-days trends (Fig 1, Fig 5). The larger TC location
18 differences for the mid-period correlates well with strong CI discrepancies and opposite TC-
19 days trends. In the last decade both TC location and CI differences show downward
20 tendencies. TC activity trends in that time are similar for JTWC and JMADT, even for the
21 strongest categories.

22 3.4 Additional contributors for BTD inconsistencies

23 The analysis shows that differences in CI numbers and TC locations share a strong
24 relationship. They are most distinguishable in the years 1987-1998, when the aircraft
25 reconnaissance terminated and development of the intense satellite measurements began. Such

1 coincidence suggests the usage of different information sources by JTWC and JMA may be a
2 reason for the given TC trend differences. JMA reports usage of geostationary imageries only
3 as a source for intensity estimation. In contrast, JTWC's (Velden, 2004) operational center
4 uses all available satellite data to ascertain the location and underlying storm structure and
5 therefore improves the information used for imagery processing with the Dvorak technique.
6 Such practises in JTWC might increase intensity values and contribute strongly to increasing
7 tendencies of intense TC-days.

8 Increasing coverage of microwave observations (SSM/I) from 1987 onwards which
9 reached the maximum in 1997, together with high-resolution scatterometer (ERS-2) measuring
10 in 1995-1997, helped in TC center positioning and analysis of the lower intensity systems.
11 Enhanced radar usability and additional information of higher-resolution TRMM in 1997
12 improved the accuracy of Dvorak-based estimations in JTWC. Introducing more and better
13 spatially-resolved data certainly could affect the data set homogeneity and statistical
14 information concerning derived trends. Extensive and irregular use of additional
15 supplementary sources by one operational center and not the other, might lead to large CI
16 discrepancies and opposite trends of intense TCs activity in comparison to other BTDC. The
17 strong, increasing tendency in intense TC-days found in JTWC, especially for the period
18 1987-1999, might be severely biased by inhomogeneities introduced by changing procedures
19 and different information sources applied in the operational centers.

20 We suggest that apart from differing methods for converting CI numbers to intensities,
21 CI discrepancies are the main contributor to differences between TC activity trends. Our
22 analysis indicates that discrepancies among operational parameters occur due to different data
23 used as input for the Dvorak method applied in JTWC. However, to check the credibility of
24 these parameters, they need to be compared with reference data.

3.5 Can the CI discrepancies between BTD sets be evaluated? A NOAA-BTD, aircraft-BTD comparison

To evaluate CI discrepancies, records for the years 2000-2008 in NOAA, JTWC and JMADT were analysed for the TS category. The main input of NOAA, QuikSCAT is stated as having highly reliable values for moderate and high TS values, while slightly overestimating wind of tropical depression strength. However, it provides data adhered to 8-min average. For this reason, NOAA can underestimate values up to 2 m s^{-1} when comparing with 1-min wind speed values within TS category.

Figure 7 presents computed annual mean differences for concurrent records between JTWC and JMADT. In this comparison JTWC reveals systematically higher values compared to JMADT. For less than 15% of all cases the absolute difference is smaller than 2 m s^{-1} which, according to Kruk et al. (2011), is within the range of the remapping method's accuracy. However, for the majority of cases (60%) JTWC is higher than JMADT by 2-8 m s^{-1} . For our comparison the data was divided into two groups according to these relationships. For the first one, representing almost 60% of cases, JMADT remains like JTWC within the TS category. For the second group, representing over 40% of the cases, JMADT is low enough to fall into the TD category. To assess which agency gives more reliable parameters, these two groups are compared with NOAA. They are analysed separately, with a greater focus on the first one (TS) due to high reference data reliability.

Figure 8a,b presents mean differences for NOAA minus JMADT and for NOAA minus JTWC, computed for the whole analysed period, for both groups. For the group that contains data of both analysed BTD within the TS category, NOAA remains closer to JMADT with 26% of the records remaining within absolute difference of 2 m s^{-1} and 50% within 4 m s^{-1} . However, NOAA presents slightly higher values than JMADT with a median for the

1 differences in the range $<0;2> \text{ m s}^{-1}$. In comparison with JTWC, NOAA has lower values for
2 more then 60% of the records, with the median within the range of $<-4;-2> \text{ m s}^{-1}$.
3 For the second group, where JMADT indicates the TD phase, only 15% of the NOAA values
4 remain within absolute difference of 2 m s^{-1} of JMADT. Here NOAA presents stronger
5 tendencies towards higher values with a median of the difference in the range of $<4;6> \text{ m s}^{-1}$.
6 However, this might be caused by a positive bias introduced by scatterometer data during
7 rainy conditions for tropical depressions. Despite this fact, JTWC still remains higher than
8 NOAA in almost 50% of the cases. Figure 9a presents a TC from 2008 where JTWC wind
9 values were higher during the whole event, except for the TD and early TS phase when
10 NOAA showed the highest values. For this TC the NOAA values remained noticeably closer
11 to JMADT.

12 The highest discrepancies still remain in the highest wind categories, therefore an
13 evaluation of adjustment methods for categories 4-5 is crucial for determining trends in TC
14 activity. For two intense TCs, Jangmi in 2008 and Megi in 2010, aircraft measured maximum
15 sustained wind speed are available. For the TC Jangmi maximum wind speed estimates of
16 JTWC (72 m s^{-1}) match the observed ones given by SMFR better than JMADT. For this case
17 JMADT presents the highest values (79 m s^{-1}), while SMFR 60-sec observations show 68 m
18 s^{-1} . Fig 9c also shows the supertyphoon Megi in 2010 for which maximum wind speed was
19 measured during an aircraft campaign as well. For this event, SFMR measurements, even
20 after averaging by 60 s interval, show the highest values (90 m s^{-1}), 87 m s^{-1} for JMADT and
21 82 m s^{-1} for JTWC.

22

23 3.6 Accuracy of intensity estimations given by BTD sets

24 To evaluate CI discrepancies, BTD records were compared with satellite-based NOAA
25 data and aircraft observations. NOAA serves as reference data for the lower intensity
26 categories, while aircraft observations are used for the highest wind speed evaluation.

1 Wind values derived from NOAA that for the TS phase, provides data with reliable
2 accuracy, remain closer to JMADT than JTWC. Nevertheless, still a wide spread of
3 differences exists among the data. JTWC shows much higher values than NOAA and
4 JMADT, even in the group where JMADT falls into the TD category and a possible positive
5 bias in NOAA has been taken into account. This indicates possible intensity overestimations
6 in JTWC due to an erroneous contribution of CI parameters. Such overestimations may also
7 be caused by supplementary data usage of JTWC, e.g. QuikSCAT, which gives values
8 averaged over an 25 km area and an 8-minute interval. These values would be treated as the
9 minimum threshold for estimated by a forecaster maximum wind speed. In the result, JTWC
10 may increase the final wind estimates to compensate for possible underestimations due to
11 wind retrieval limitations. Figure 9a shows time series of TC intensity for typhoon Dolphin in
12 2008 and serves as an example for pronouncedly higher wind speed values of JTWC in
13 comparison to reference data (NOAA) and alternative BTD. However, the indirect way of
14 choosing the maximum wind speed for NOAA winds (which provide reliable information
15 only for lower TC intensity categories), as well as the limited accuracy of the remapping
16 method still contribute to the uncertainty in our estimation of BTD reliability.

17 We now focus on CI parameters in the higher part of the SSHS intensity scale, where
18 the strongest discrepancies still remain. An evaluation of BTD for categories 4-5 is crucial for
19 determining trends in TC activity. As aircraft sensors are unable to provide direct
20 measurements of 10 m 1min sustained wind speed, they serve only as input to prepare surface
21 wind analyses. Here the initialization conditions and assimilation techniques are crucial to
22 construct reliable analyses. Figure 9b presents BTD, aircraft observations provided by SFMR
23 taken during the TCS-08 2008 campaign, and an analysis reconstructed with those
24 observations (Zhang et al., 2007) for typhoon Sinlaku in 2008. The initialization scheme
25 assimilates TC central minimum pressure given by JTWC, but the maximum wind speed for
26 higher categories does not reach JTWC values. As the provided TC reconstruction may be

1 also biased due to the 10-km horizontal resolution, this can complicate the evaluation of BTD.
2 On the other hand, the JTWC report (JTWC, 2009) states, that the aircraft measurements
3 themselves for this TC had decisive impact on intensity estimation. Aircraft reconnaissance in
4 this case helped to identify the second intensification phase. While for the first intensification
5 phase the Dvorak technique estimated intensity with good accuracy, it underestimated the TC
6 intensity during the second phase. The reconstructed reanalysis for the second period matches
7 the observed values.

8 For typhoon Jangmi the flights during the TPARC aircraft campaign occurred at the
9 time of TC maximum intensity for which a mean 60-sec value of 68 m s^{-1} was measured
10 while JMADT estimated the highest values (79 m s^{-1}). For Megi, SFMR 60-sec
11 measurements show the highest values (90 m s^{-1}) of maximum wind speed. Additionally, the
12 SFMR recorded the weakening of TC Megi faster than estimated by the Dvorak method.
13 Landfalling TC situations, for which the reliability of Dvorak relationships is limited, require
14 in-situ observations. Nakazawa and Hoshino (2009) also noticed differences in operational
15 (CI- and T-) numbers among various BTD, both for intensification and weakening phases.
16 Differing weakening ratios, after reaching TC maximum intensity in BTD sets, indicate that
17 there may be differences between definitions for allowable intensity change (in the form of CI
18 and T parameters). Such constraints (Dvorak,1984) were gradually relaxed by JTWC during
19 the 1990s (Dvorak, 2004), allowing for a faster weakening of intense TCs. These procedural
20 changes possibly contributed to the existing discrepancies among BTD.

21 Additionally, it is noticeable that in the developing stage of a typhoon, BTD in JTWC
22 is strongly influenced by aircraft measurements (Figure 9c). These were possibly used to
23 supplementary identify the early intensification phase.

4.1 Summary and Conclusions

2 This paper assesses the reliability of BTD in climate statistics for the WNP region.
3 We confirmed that the different methodologies to derive TC intensities used by the
4 meteorological agencies producing BTD influence TC activity trends. Therefore the skill of
5 methods to minimize discrepancies between the individual data sets was evaluated. Both the
6 commonly used linear factor multiplication method (used to homogenize BTD with different
7 wind speed intervals) as well as the method of Knapp and Kruk (2010) show high skill to
8 reduce trend discrepancies, but only when categories 2-5 are considered together. Then all
9 BTD show increasing numbers of annually accumulated TC-days for the period 1977-2008.
10 However, when analysing categories 2-3 and 4-5 separately, the methods' skills differ. We
11 found that using a multiplication factor lead to overestimated trends of TC-days for lower
12 categories (2-3) while still underestimating the highest ones (4-5).

13 An alternative method, which reconstructs TC intensity by remapping CI parameters
14 with a DT conversion (Knapp and Kruk, 2010) reduces most discrepancies for categories 2-3.
15 For the highest categories, the technique minimizes discrepancies only partly, TC activity
16 trends in JMADT show no trend while strongly increasing trends are visible for JTWC.

17 The application of the same converting procedures to retrieve TC intensities should
18 theoretically reduce the difference between the individual BTD to zero. However, remaining
19 differences indicate that there are additional contributing factors leading to discrepancies in
20 operational CI numbers.

21 The distribution of the CI discrepancies in time corresponds to the differences in TC
22 center positions. The largest discrepancies occur in the 1980s when higher-resolution satellite
23 observations were developing. Toward the latter half of the decade, the reduction and phasing
24 out of aircraft data sources may also have had an influence.

1 This indicates that extensive and irregular use of additional supplementary sources by JTWC
2 might cause huge CI discrepancies and opposite trends of intense TCs activity with other
3 BTD. The strong increasing tendency in intense TC-days found in JTWC, especially for the
4 period 1987-1999, might be severely biased by inhomogeneities introduced by changing
5 procedures and information sources. Using only the geostationary satellite imageries for
6 intensity estimations by JMA limits its accuracy. On the other hand, this maintains
7 homogeneity within the data set which makes this source more reliable for deriving climate
8 statistics.

9 CI numbers and wind intensity of JTWC and JMADT were compared to NOAA sea
10 surface wind speeds and aircraft measurements to evaluate which BTD provides more
11 accurate estimations. JTWC shows a systematic overestimation of both NOAA and JMADT
12 for the TS category, where NOAA data is considered to be very accurate. For the TS category
13 JMADT wind speed values remain closer to NOAA, although visible differences still exist.
14 Higher CI parameter estimates as well as subjective interpretation of additional sources in
15 JTWC (e.g. microwave wind retrievals) likely contribute to such results. We conclude that
16 JMA provides more reliable CI parameters than JTWC for the TS wind speed range.

17 Sparse in-situ data limit the evaluation effort of BTD accuracy for the highest wind
18 regimes. Aircraft campaign measurements in 2008 and 2010 show some agreement with
19 maximum intensity estimations in BTD. For a more complete evaluation, aircraft data for the
20 earlier period would be needed, when the accuracy uncertainties were the highest.

21 Additionally, the analysis of some strong TC events like Sinlaku (2008) and Megi
22 (2010) suggests that there are some deficiencies within the Dvorak technique procedures.
23 Slow weakening ratios for BTD in comparison to in-situ observations indicate that not only
24 the homogeneity has to be assured, but also that temporal non-changing methods to estimate
25 TC intensities should be applied in all operational centers.

We emphasize the importance to document those operational procedures that are applied for the Dvorak technique by the meteorological agencies, otherwise the interpretation of the results can lead to misleading conclusions. This may happen when considering ambiguously specified wind speed definitions in CMA or intensity in JMA before applying Koba et al. (1991) conversion table. We suggest paying special attention with regard to the highest wind regimes as the largest differences between BTD sets were found here. The differences in TC activity trends may require academic agreement on a set of procedures and a reanalysis of existing storm data.

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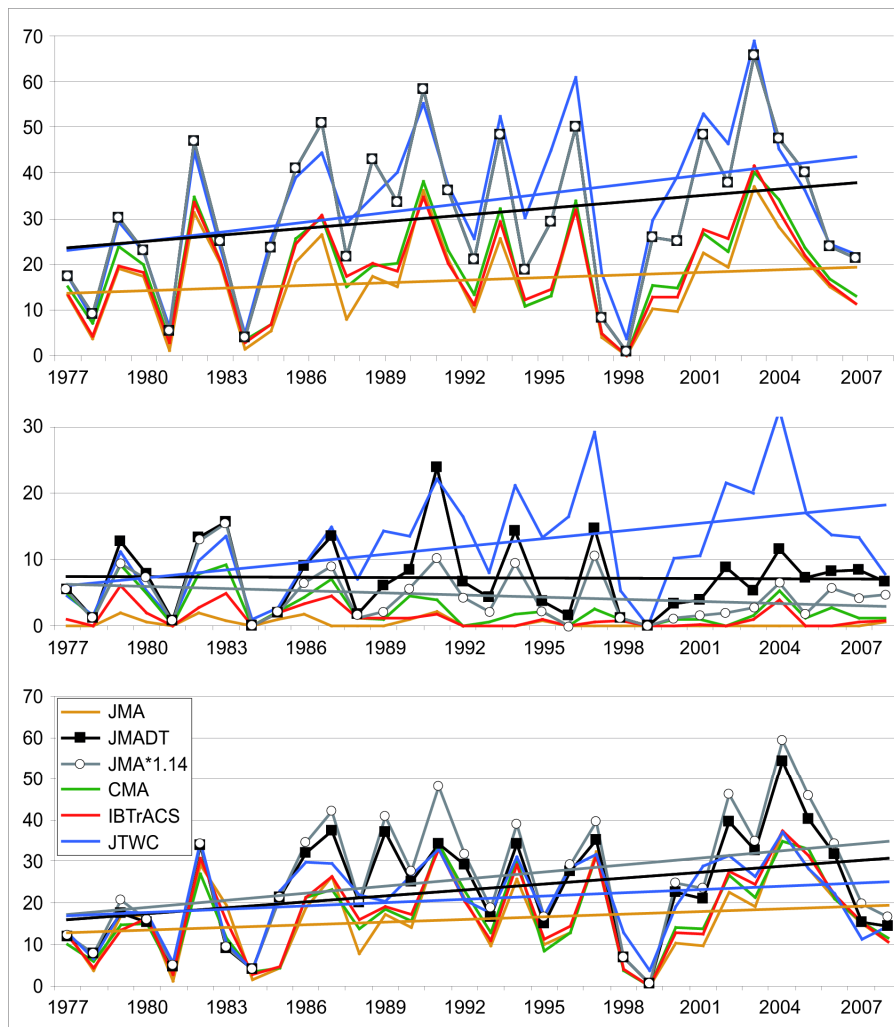


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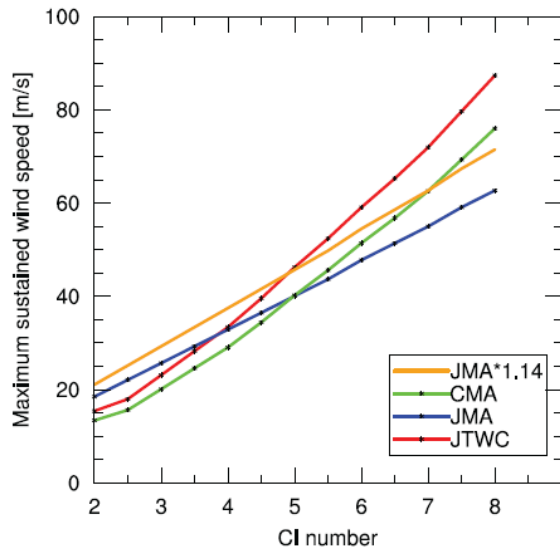


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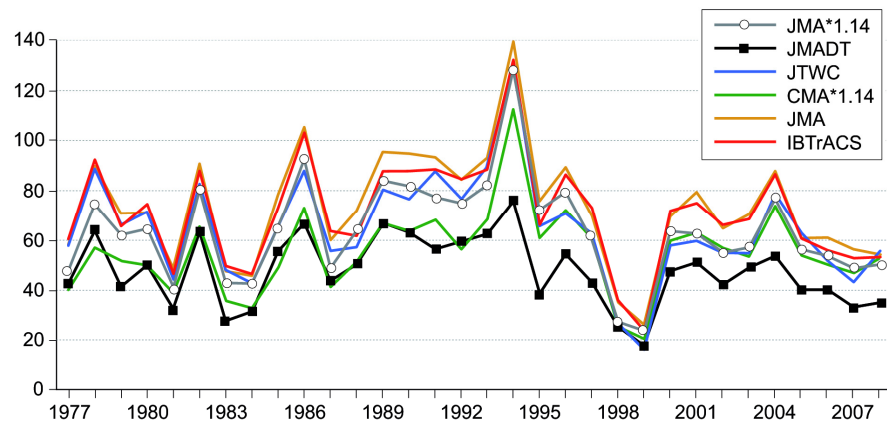
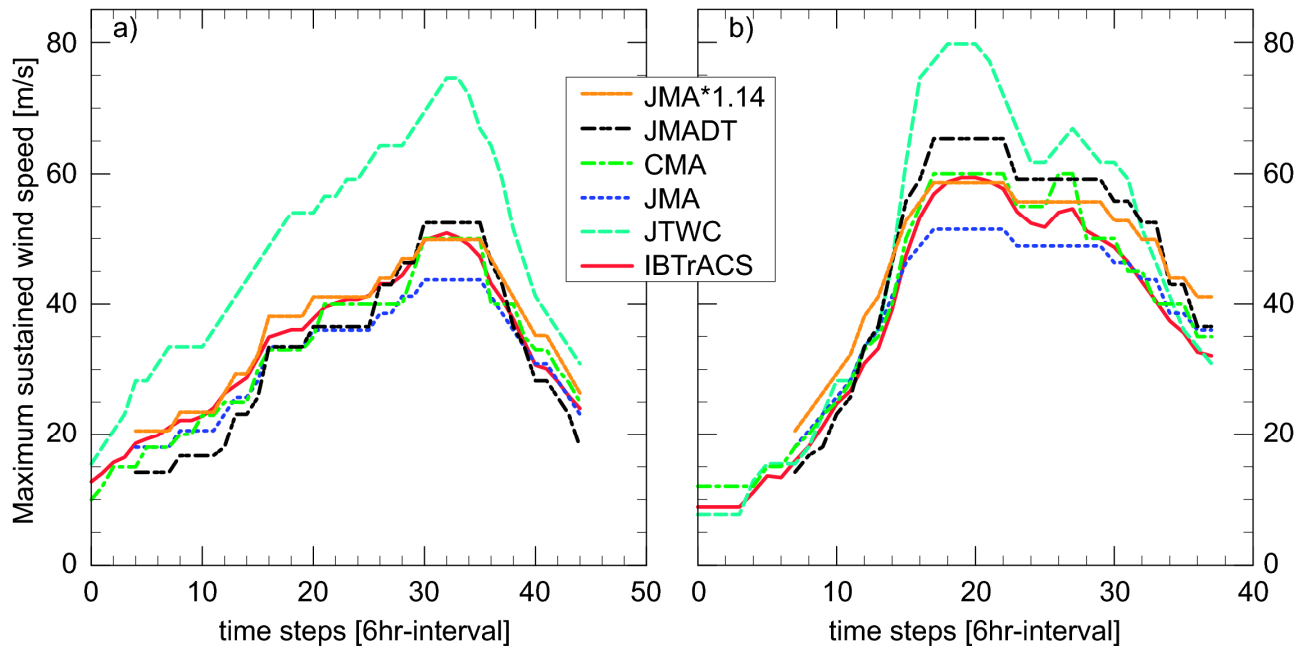


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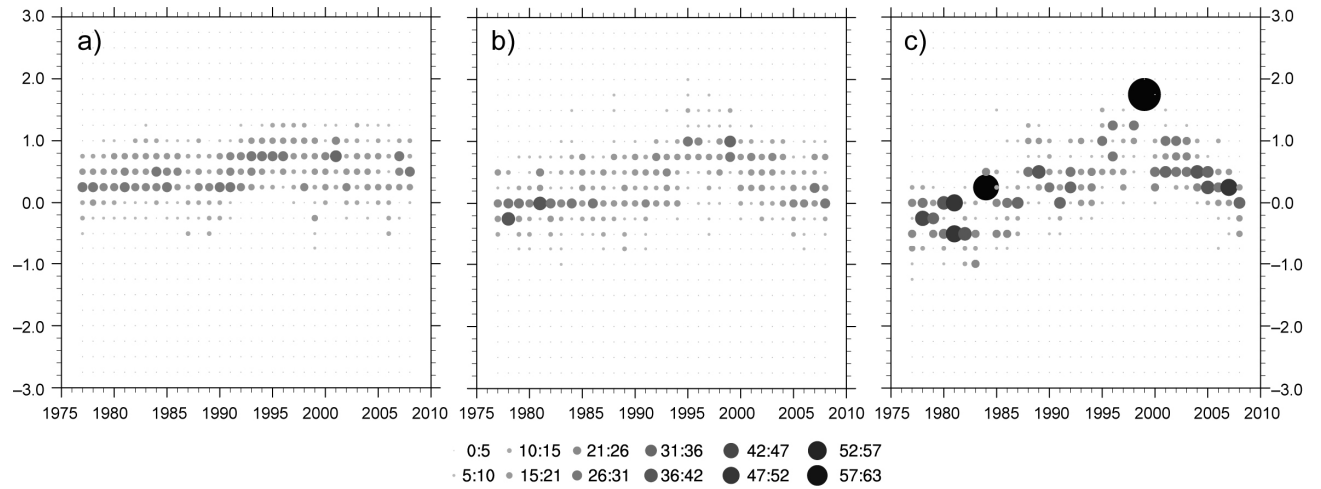


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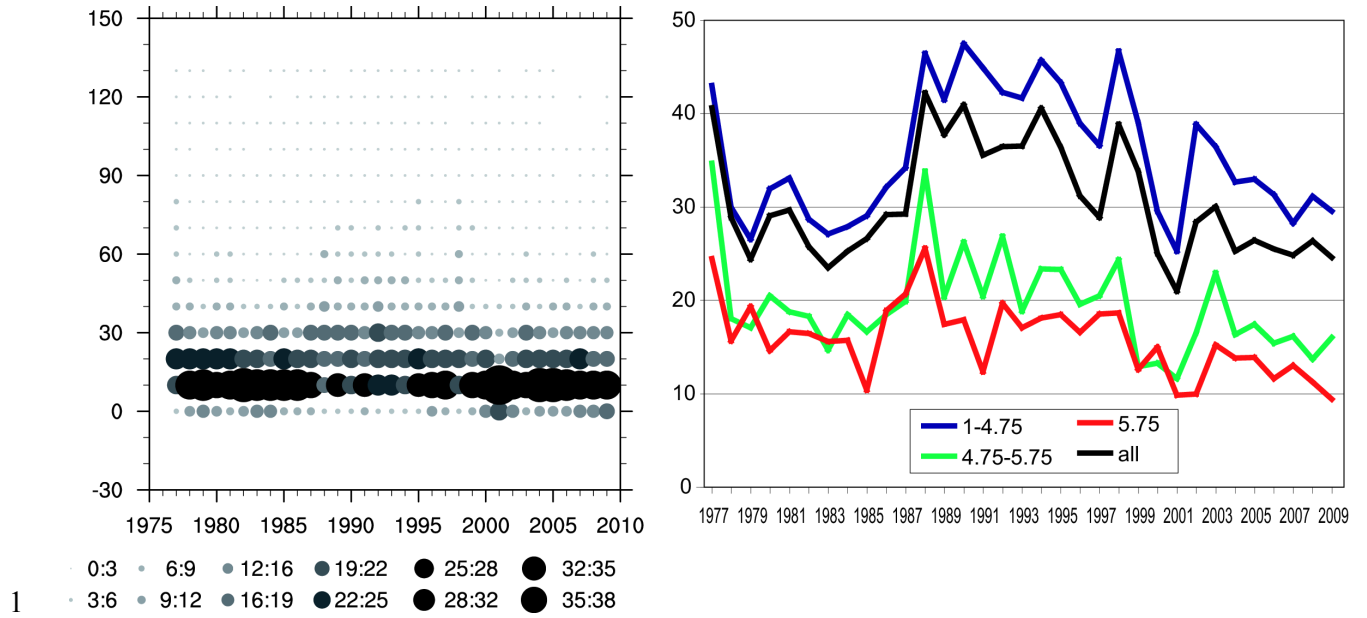


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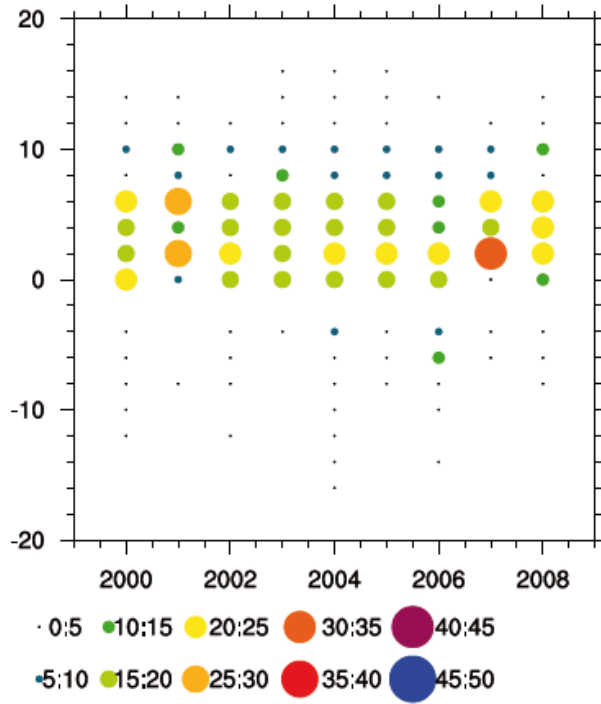
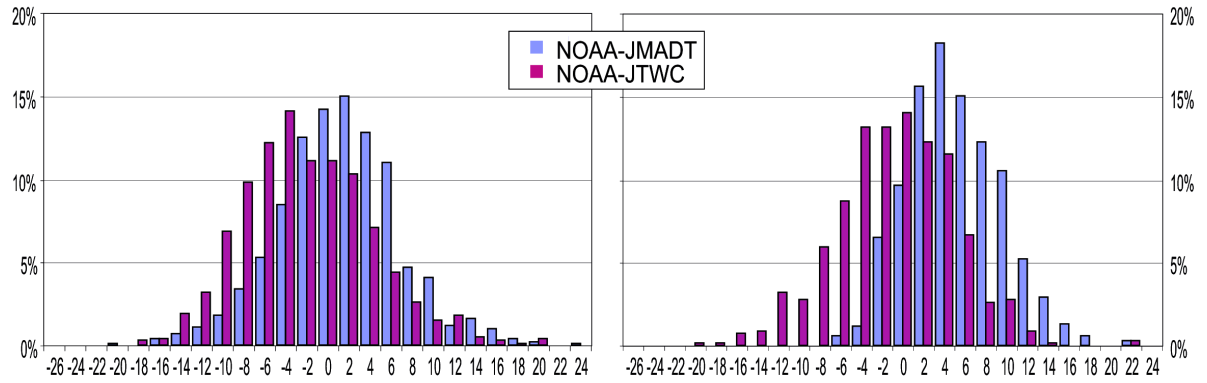


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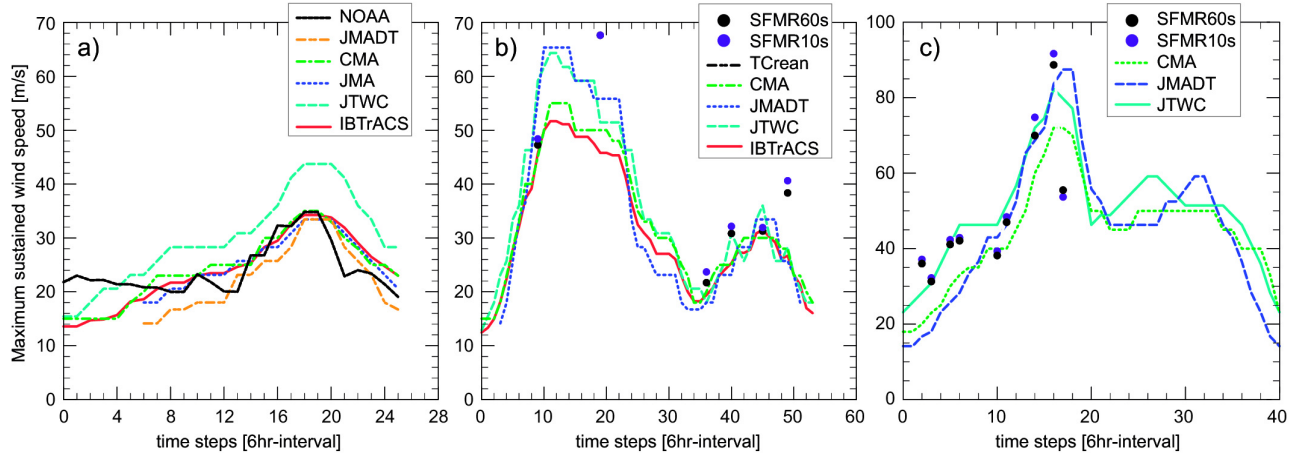
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