

Variability of Tropical Cyclone Representation within Atmospheric Reanalysis Datasets Ben Schenkel (bschenkel@fsu.edu) and Robert Hart (rhart@fsu.edu)





Introduction

The emergence of reanalyses has provided tools of great utility for studying interactions between tropical cyclones (TCs) and their larger scale environment. In spite of the increasing usage of reanalyses for climate scale studies of TCs, there has been no comprehensive examination of the representation of TCs within these datasets. The implications of the accurate depiction of TCs within reanalyses may have far reaching consequences including potentially impacting the representation of the general circulation on short time scales. The following study seeks to quantitatively compare reanalysis TC position, intensity, and structure with observational datasets and examine how these parameters vary within and amone

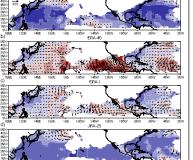
Methodology

reanalysis datasets.

In this study, the fidelity of TC position, intensity, and structure is examined within five reanalyses: the NCEP CFSR (Saha et al. 2010), the ECMWF ERA-40 (Uppala et al. 2005), the ECMWF ERA-I (Simmons et al. 2007), the JMA JRA-25 (Onogi et al. 2007), and the NASA MERRA (Bosilovich et al. 2006). TCs that are found equatorwards of 36°N within the NHC best-track dataset (Jarvinen et al. 1984: Neumann et al. 1993) and JTWC best-track dataset (Chu et al. 2002) in the eastern North Pacific. North Atlantic, and western North Pacific between 1979-2001 are chosen for study Each best-track TC within the reanalysis is manually tracked using minimum mean sea-level pressure and maximum 925 hPa relative vorticity. The reanalysis position and intensity are then compared to those found in the best-track dataset.

In addition to examining these traditional metrics, parameters from the cyclone phase space (Hart et al. 2003) and storm-relative composites of anomalies are used to evaluate TC structure. The composites are constructed by bilinearly interpolating each grid to a uniform horizontal resolution centered about the manually tracked reanalysis TC position. Each reanalysis TC is then composited into one of four intensity bins according to its best-track intensity.

Spatial Variation of Tropical Cyclone Position Difference



140W 120W

Figure 1: Plan view of the mean position difference (km) magnitude (shaded) and vector displacement (army, reamplayin point) nethiev to best-track) for the (a CFSR, (b) ERA-l, (d) RA-25, and (e) MERRA for TCs in the eastern North Pacific, North Altantic, and western North Pacific. The mean magnitude of the position differences with product to a 2° by 2° grid representing the average of the position difference weighted by its distance from the gridpoint. Vectors excluded for gridpoint with values of less than 100 km. The grid is amonoted once with a 100 km. The gri

160E

nine-point smoother

Mean Tropical Cyclone Position and Intensity Differences

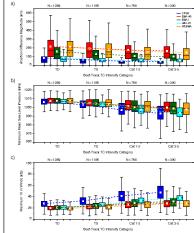


Figure 2: Box and whiskers plot of (a) position difference magnitude (tax), (b) minimum mean scheep resoure (liph), and (c) maximum (10) with (a) they for the sears PostP Pacific, (North Atlantic, and western North Pacific (exch of the five reanalyses startfield by the four bost-track intensity and western North Pacific (exch of the five reanalyses startfield by the five bost-track intensity color coding of the color of the color

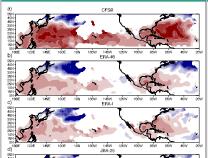


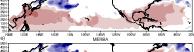
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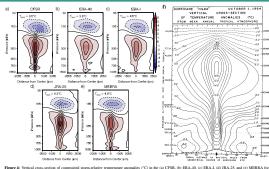
Spatial Variation of Low-Level Tropical Cyclone Structure







North Atlantic Category 3-5 Tropical Cyclone Structure in Reanalyses and Observations



North Atlantic category 3-5 TCs. (f) Vertical cross-section of storm-relative temperature anomalies for TC Hilda on October 1, 1964 (Hawkins and Rubsam 1968).

Discussion

The results of this study show that reanalysis TC representation exhibits significant variability among basins. Specifically, North Atlantic TCs have the smallest position differences, strongest intensities, and largest composited anomaly magnitudes. The robust representation of North Atlantic TCs is likely attributable to a relatively greater density of observations. In contrast, eastern North Pacific TCs have the largest position differences, weakest intensities, and smallest composited anomaly magnitudes. The poorer representation of eastern North Pacific TCs is possibly due to the smaller mean size of TCs relative to other basins (Chavas and Emanuel 2010; Schenkel and Hart 2011) and a relative dearth of observations (Hatsushika et al. 2006; Manning 2007; Vecchi and Knutson 2008). Artificial trends in reanalysis TC representation are also seen likely due to variations in the density of observations, observed TC age, observed TC size, and the resolution of the reanalysis. Among the reanalyses, the most realistic representation of TCs is found within the CFSR and JRA-25 due to the use of vortex relocation (Liu et al. 2000) and TC wind profile retrievals (Hatsushika et al. 2006), respectively. While the coarse resolution of reanalyses precludes the replication of the observed intensity of TCs, the gross structure of TCs is captured reasonably well. Given the particularly poor representation of the strongest intensity TCs, caution should be exercised when utilizing these datasets for calculations dependent on resolving the absolute magnitude of quantities such as TC intensity metrics.

Acknowledgments and References

This research is supported by NASA Headquarters under the NASA Farth and Space Science Fellowship Program Grant 10. Earth10R-0036 and NSF Grant ATM-0842618. The authors also would like to thank Michael Bosilovich (NASA GSFC) for fruitful discussions regarding MERRA. We acknowledge the FSU HPC facility and staff for contributions to the results presented in this poster.

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