

ABSTRACT

The 2017 Hurricanes season was one of the most powerful severe weather events producing catastrophic socio-economic and environmental effects on the east coast of the United States. Therefore, tracking their path accurately is extremely useful. Today Global Navigation Satellite Systems (GNSS) tropospheric products, such as Zenith Wet Delays (ZWD), and Integrated Water Vapor (IWV) are used as complementary data sets in Numerical Weather Prediction (NWP) models. In this study, we employed GPS-derived IWV and horizontal tropospheric gradient information to monitor and investigate the complicated characteristics of hurricane events in their spatial and temporal distribution using a dense ground network of GPS stations. Our results show that a surge in GPS-derived IWV occurred several hours prior to the manifestation of the major hurricanes Harvey and Irma. We used the derived GPS-derived IWV information as input to spaghetti lines weather models, allowing us to predict the paths of Harvey and Irma hurricanes. As such, a parameter directly estimated from GPS can provide an additional resource for improving the monitoring of hurricane paths.

INTRODUCTION

Extreme meteorological events like hurricanes Harvey and Irma emphasised the need for tracking using additional observational tools to mitigate the risks. GPS tropospheric products can provide useful information to understand the structure and behavior of hurricanes better[1]. The use of GPS tropospheric products for regional severe storm prediction is an application that the geodetic community wishes to promote to meteorological research groups. The main motivation for this work is to track and characterize hurricane movements using GPS-derived local maxima IWV patterns. We introduce an emprical spaghetti model [2] using GPS-derived atmospheric water vapor fields and how these fields can potentially predict the path of hurricanes. From maps of IWV fields, we have constructed a pair of geodetic coordinate representing GPS-derived IWV local-maxima every 6-hour interval. Then, we employed the spaghetti model to provide early information on where a potential tropical storm or hurricane might occur. We have also compared the temporal distribution of the GPS-derived IWV content with the precipitation value from Tropical Rainfall Measuring Mission (TRMM) and Integrated Multi-Satellite Retrievals for Global Precipitation Measurement (GPM/IMERG) satellite mission.

Data and Methods

Retrieval of IWV: We have used the zenith total delay (ZTD) obtained at the Nevada Geodetic Laboratory (NGL)[3] using GIPSY/OASIS-II software in a precise point positioning (PPP) strategy [5] for a dense array of GPS stations shown in Fig. 1a. We have calculated IWV by subtracting ZHD (from VMF grided files) from the ZTD using the formula

$$IWV = \frac{(ZTD - ZHD)}{10^{-6}(K'_2 + \frac{K_3}{T_m})R_v\rho}$$
(1)

where K_2 and K_3 are reflectivity constants, Rv is the ideal gas constant, ρ is the density of the water vapor, and T_m is the atmospheric column mean temperature. T are derived from the ECMWF.

Spaghetti model: To track the movement of Harvey and Irma, we constructed the GPS-IWV map for every 6 hour. Then, we extracted the local-maxima of GPS-IWV from the maps, estimating at a location (lon_o, lat_o) of a zero crossing of the first derivative

$$\frac{\partial IWV(lon_o, lat_o)}{\partial lon}\Big|_{t=6hr \ priori} = \frac{\partial IWV(lon_o, lat_o)}{\partial lat}\Big|_{t=6hr \ priori} = 0, \qquad (2)$$

$$IWV_{max} = IWV(lon_o, lat_o) . (3)$$

of the IWV field, $IWV \propto f(lon, lat, t = 6 hr)$. superimposed on daily GPM/IMERG precipitation (blue-line) time series for the two where Further, we employed the second-derivative to determine a local-maxima IWV and its coordinates. One could visualize each line in this model to be a single strand of spaghett that is formed by connecting the individual local maximum. To introduce sufficient variation, we added white noise to the local IWV maxima. where r is geodetic coordinate of IWV_{max} and ξ is assumed to Gaussian distributed the estimated linear regression. The number of data samples (N), Pearson correlation, noise with a certain mean and variance. indicate the data in nighttime while the magenta dots indicate the date in daytime.

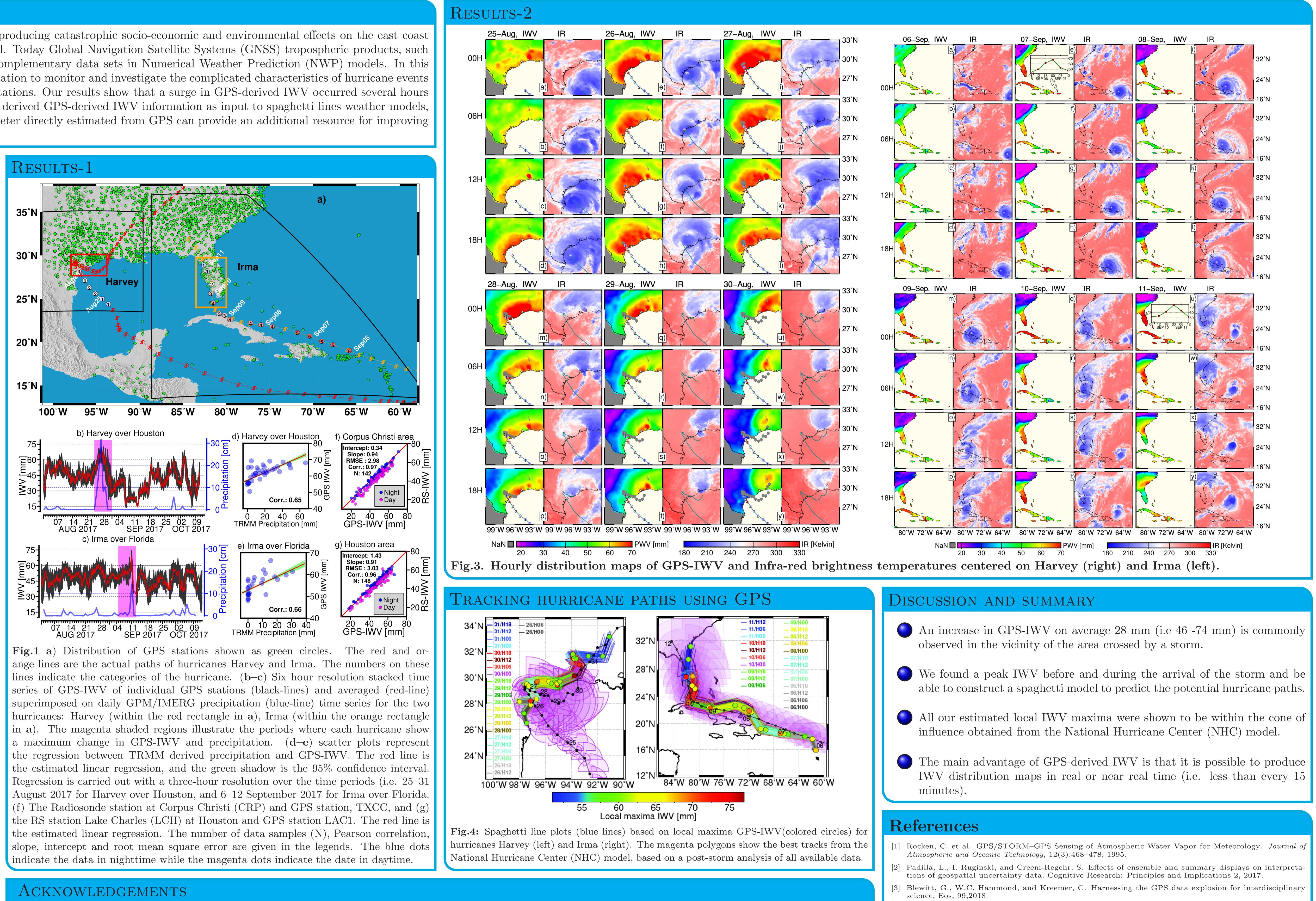
$$f \sim f(lon, lat) = r(lon_o, lat_o) + \xi \tag{4}$$

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Tracking hurricanes Harvey and Irma using GPS tropospheric products

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- able to construct a spaghetti model to predict the potential hurricane paths.
- C The main advantage of GPS-derived IWV is that it is possible to produce IWV distribution maps in real or near real time (i.e. less than every 15

- Rocken, C. et al. GPS/STORM-GPS Sensing of Atmospheric Water Vapor for Meteorology. Journal of
- Huffman, G.J., D. Bolvin, and Nelkin E.J. Integrated MultisatellitE Retrievals for GPM (IMERG) Tech-
- nical Documentation NASA, pp:1-46, 2017. Zumberge, J., V. B. Hein, M., Jeerson, D., Watkins, M., and Webb, F. Precise Point Positioning for the Efficient And Robust Analysis of GPS Data from Large Networks. JGR, 102, 1997.