



1 **Capturing the signature of heavy rainfall events using the 2-d/-4-d**
2 **water vapour information derived from GNSS measurement**
3 **in Hong Kong**

4
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11 **Abstract:** Apart from the well-known applications like positioning, navigation and timing (PNT),
12 Global Navigation Satellite System (GNSS) has manifested its ability in many other areas that are
13 vital to society largely. With the dense setting of the regional continuously operating reference
14 station (CORS) networks, monitoring the variations in atmospheric water vapour using a GNSS
15 technique has become the focus in the field of GNSS meteorology. Most previous studies mainly
16 concentrate on the analysis of relationship between the two-dimensional (2-d) Precipitable Water
17 Vapour (PWV) and rainfall while the four-dimensional (4-d) variations of atmospheric water
18 vapour derived from the GNSS tomographic technique during rainfall events are rarely discussed.
19 This becomes the focus of this work, which investigates the emerging field of GNSS technology
20 for monitoring changes in atmospheric water vapour during rainfall, especially in the vertical
21 direction. This paper includes an analysis of both 2-d, and 4-d, precipitable water vapour profiles.
22 A period with heavy rainfall events in this study was selected to capture the signature of
23 atmospheric water vapour variation using the ground-based GNSS tomographic technique. GNSS
24 observations from the CORS network of Hong Kong were used. Analysed results of the 2-d
25 PWV/4-d water vapour profiles change during the arrival, occurrence, and depression of heavy
26 rainfall show that: (i) the PWV time series shows an increasing trend before the arrival of heavy
27 rainfall and decreases to its average value after the depression of rainfall; (ii) rainfall leads to an
28 anomalous variation in relative humidity and temperature while their trends are totally opposite
29 and show daily periodicity for periods without rain (this is highly correlated with the changes in
30 solar radiation); (iii) atmospheric water vapour presents unstable conditions with intense vertical



31 convective motion and hydrometeors are formed before the arrival of rainfall while returning to
32 relatively stable conditions during heavy rainfall. This study indicates the potential for using
33 GNSS-derived 2-d PWV and 4-d profiles to monitor spatio-temporal variations in atmospheric
34 water vapour during rainfall, which provides a better understanding of the mechanism of
35 convection and rainfall induced by the extreme weather events.

36 **Keywords:** GNSS; PWV; water vapor profiles; extreme weather events

37

38 **1 Introduction**

39 Precipitable water vapour (PWV), which refers to the total content of integrated water vapour
40 density along the zenith direction, is a significant component reflecting the short-term atmospheric
41 water vapour variations used in severe weather detection as well as in long-term climate studies
42 (Bai, 2004; Liu et al., 2013); however, it is difficult to obtain a satisfactory spatio-temporal
43 resolution of atmospheric water vapour due to the limitation of both the number of traditional
44 sounding stations and the observation times (Brenot et al., 2013; Zhang et al., 2015). For the past
45 20 years, the ability to estimate water vapour contents with an accuracy of 1 to 2 mm has been
46 proved using the Global Navigation Satellite System (GNSS), which generally formed a new field
47 of study in GNSS Meteorology (Bevis et al., 1992). Therefore, the variation of atmospheric water
48 vapour with high accuracy, as well as the high spatio-temporal resolution can be obtained using
49 the hyper-dense GNSS networks (with receivers only a few kilometres apart).

50 PWV, at high spatio-temporal resolution is an indicator for monitoring the water vapour responses
51 to severe weather events (Zhang et al., 2015; Yao et al., 2017; Zhao et al., 2018a, 2018b). It has
52 been used for operational meteorology in some areas such as Japan (JMA, 2013), the UK (Bennitt
53 and Jupp, 2012), France (Guerova et al., 2016), and Italy (Barindelli et al., 2018). In those areas,
54 the zenith total delays (ZTD) or PWV estimated from ground-based GNSS measurements are
55 generally assimilated into numerical weather prediction (NWP) models (De Haan 2013; Saito et
56 al., 2017). In addition, ZTD or PWV is also used for the early warning and forecasting of severe
57 precipitation, which has been investigated in areas of Greater Lisbon in Portugal as well as
58 Zhejiang Province in China (Benevides et al., 2015; Yao et al., 2018; Zhao et al., 2018a, 2018b).
59 These applications have verified the ability of GNSS as used in meteorology, but those cases are
60 mainly focussed on two-dimensional (2-d) PWV which cannot reflect the specific vertical



61 variations in atmospheric water vapour.

62 Although GNSS tropospheric tomography has been proposed (Flores et al., 2000), and can be used
63 to obtain four-dimensional (4-d) water vapour variations, the development of this technique has
64 mainly focussed on improvement of theoretical and model aspects while its application is rarely
65 discussed. For example, the reliability of GNSS tomography was validated using radiosonde data
66 by Seko et al. (2000) and Troller et al. (2002, 2006). The joint reconstruction of atmospheric water
67 vapour was also investigated by combing multi-GNSS observations as well as multi-source data
68 derived from the Constellation Observing System for Meteorology, Ionosphere, and Climate
69 (COSMIC), Interferometric Synthetic Aperture Radar (InSAR), radiosonde flights, *etc.* (Bender
70 and Raabe, 2007; Bender et al., 2011; Wang et al., 2014; Alshawaf, 2013; Heublein et al., 2015;
71 Benevides et al., 2015; Zhao et al., 2018c). For the improvement of tomographic models and
72 resolution thereof, Perler et al. (2011) proposed a new parameterised tomographic method, which
73 is capable of obtaining better tomographic results. Some methods concerned with the resolution of
74 tomographic models, as well as the division of tomography areas, have been proposed such as the
75 extended sequential successive filtering method, iterative reconstruction algorithm, *etc.* (Braun et
76 al, 2003, 2004; Wang et al., 2014; Zhao et al., 2017a, 2018d; Chen and Liu, 2014). In addition,
77 maximal use of GNSS signals penetrating from the side faces of tomography areas has obtained a
78 significant improvement and is realised by introducing the water vapour scale factor (Yao and
79 Zhao, 2016; Yao et al., 2016; Zhao et al., 2017b).

80 Currently, GNSS tomography technique is maturing in terms of theoretical and model aspects
81 through almost 20 years of development, but its application in GNSS meteorology remains to be
82 further investigated, therefore, we focus on capturing the signature of heavy rainfall events using
83 the 2-d/4-d water vapour information derived from GNSS measurements in Hong Kong. The 2-d
84 PWV time series is first analysed for correlation with heavy rainfall. Thereafter, the signatures of
85 4-d water vapour variations derived from GNSS tomography are investigated during heavy rainfall
86 events while the tomographic modelling is resolved using the optimal weighting determination
87 method.

88

89 **2 Fundamentals of GNSS meteorology**

90 **2.1 Retrieval of GNSS PWV**



91 Satellite signals are delayed and bent when crossing the atmosphere, which adds ionosphere and
92 troposphere delay: the former delay can be eliminated based on ionosphere free (IF) linear
93 combination during the processing of GNSS measurement due to the dispersive nature of
94 ionosphere delay (Dach and Walser, 2013). The latter delay can be divided into two parts:
95 hydrostatic delay and wet delay. The first part of the tropospheric delay in a vertical direction, also
96 called zenith hydrostatic delay (ZHD), can be precisely calculated by the Saastamoinen model
97 (Saastamoinen, 1972) with the observed surface pressure while the second part can be estimated in
98 the zenith direction using GNSS data. The second part is also called zenith wet delay (ZWD),
99 from which the PWV can be calculated, thus forming a new concept: GNSS meteorology, as first
100 proposed by Bevis et al. (1992). The calculation used in obtaining PWV is expressed as follows:
101 the zenith total delay is first estimated by processing the GNSS measurements using the GNSS
102 processing software such as Bernese, GAMIT, *etc.* The ZWD is then obtained by extracting the
103 ZHD from ZTD and thus the PWV can be calculated based on the following equations
104 (Saastamoinen, 1972; Askne and Nordius, 1987; Bevis et al., 1992):

$$\begin{aligned} \text{PWV} &= \Pi \cdot \text{ZWD} \\ \Pi &= 10^6 / \left((k_2' + k_3 / T_m) \cdot R_v \cdot \rho_w \right) \\ \text{ZWD} &= \text{ZTD} - \text{ZHD} \\ \text{ZHD} &= \frac{0.002277 \times P}{1 - 0.00266 \times \cos(2\varphi) - 0.00028 \times H} \end{aligned} \quad (1)$$

106 Where Π refers to the conversion factor, where k_2' , k_3 , and R_v are constants with values of
107 22.1 K/mb, 3.739×10^5 K²/mb and 461.495 J/kg/K, respectively, T_m represents the weighted mean
108 temperature, which is related to surface parameters such as temperature and pressure. Therefore,
109 T_m is usually calculated based on the empirical model using the data from radiosonde or numerical
110 weather model due to the observed layered meteorological parameters with are rarely obtained
111 (Bevis et al., 1994; Yao et al., 2012). In the fourth formula in Eq. (1), P , H , and φ represent
112 the surface pressure (hPa), geodetic height (km), and station latitude (rad), respectively. In our
113 study, the value of T_m is calculated based on the established regional T_m model using the
114 radiosonde data and observed temperature (Section 3.2).

115

116 2.2 Establishment of tomographic model



117 Generally, the slant wet delay (SWD) or slant water vapour (SWV) is considered as the input
118 information for GNSS troposphere tomography (Flores et al., 2000; Hirahara, 2000; Skone and
119 Hoyle, 2005; Rohm and Bosy, 2009; Chen and Liu., 2014) and the following equation gives an
120 expression used to obtain SWV (Flores et al., 2000):

$$121 \quad \text{SWV}_{azi,ele} = m_w(ele) \cdot \text{PWV} + m_w(ele) \cdot \cot(ele) \cdot (G_{NS}^w \cdot \cos(azi) + G_{WE}^w \cdot \sin(azi)) \quad (2)$$

122 Where m_w presents the wet mapping function. ele and azi refer to the elevation angle and
123 azimuth angle, respectively. G_{NS}^w and G_{WE}^w are the gradient parameters in the south-north and
124 west-east directions, respectively.

125 If a sufficient number of SWVs derived from some stations in a regional CORS network can be
126 obtained, the GNSS tomographic technique can be used to reconstruct the three-dimensional (3-d)
127 distribution of atmospheric water vapour field. Therefore, a four-dimensional (4-d) water vapour
128 information is a time series of such a 3-d tomographic result, which can reflect the regional
129 atmospheric water vapour variations in both the spatial and temporal domains. As described by
130 Flores et al. (2000), the linear observation equation between SWV and water vapour density can
131 be expressed as follows:

$$132 \quad \text{SWV} = \sum (d_{ijk} \cdot x_{ijk}) \quad (3)$$

133 Where i, j, k represent the location of the area of interest in the longitudinal, latitudinal, and
134 vertical directions, respectively, d_{ijk} and x_{ijk} refer to the distance travelled by satellite signals
135 and the water vapour density remains to be estimated, respectively in the discretized voxels
136 (i, j, k) . Therefore, the matrix form of the tomographic observation equation can be described as
137 follows:

$$138 \quad \mathbf{y} = \mathbf{A} \cdot \mathbf{x} \quad (4)$$

139 Where \mathbf{y} represents the column vector of SWV derived from GNSS measurements. \mathbf{A} and \mathbf{x}
140 are the coefficient matrix of distance penetrated by satellite rays and the column vector of water
141 vapour density, respectively.

142 Due to the large sparse matrix of observation equation, some constraints are required to overcome
143 the influence caused by the ill-posed problem in the inversion of the tomographic normal equation
144 (Flores et al., 2000; Bi et al., 2006; Bender et al., 2011; Rohm and Bosy, 2011; Chen and Liu,



145 2014). In our study, both horizontal and vertical constraints are considered. The water vapour
146 density in a certain voxel is regarded as the weighted mean value of its horizontal neighbouring
147 voxels (Rius et al., 1997) and the negative exponential function is introduced to describe the
148 relationship between the nearby voxels in the vertical direction while the coefficients of functional
149 model are established using radiosonde data (Yao and Zhao, 2016). Consequently, the
150 tomographic modelling can be expressed after imposing the constraints as:

$$151 \quad \begin{pmatrix} \mathbf{y} \\ \mathbf{0} \\ \mathbf{0} \end{pmatrix} = \begin{pmatrix} \mathbf{A} \\ \mathbf{H} \\ \mathbf{V} \end{pmatrix} \cdot \mathbf{x} \quad (5)$$

152 Where \mathbf{H} and \mathbf{V} are the coefficient matrices of horizontal and vertical equations, respectively.
153 To obtain a reasonable tomographic result from the above equation, an optimal tropospheric
154 solution method is used, which can adaptively tune the weightings of different types of equations
155 (Zhao et al., 2018d).

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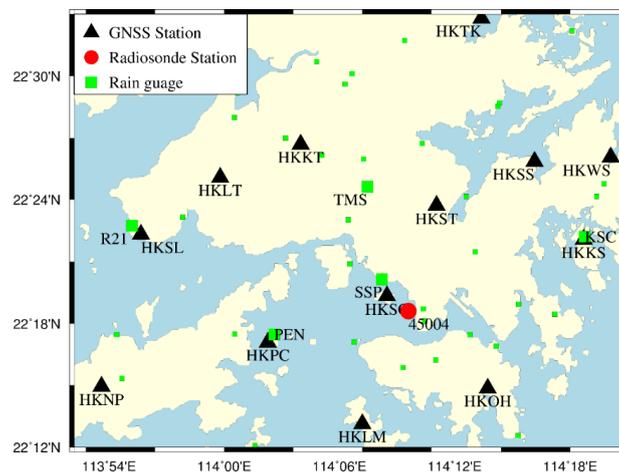
157 **3 Data description and establishment of a regional T_m model**

158 **3.1 Data description**

159 To validate the ability of GNSS technique in capturing the signature of atmospheric water vapour
160 variation during heavy rainfall events, two periods of GNSS observations (19 to 27, July 2015 and
161 1 to 8, August 2015) from 13 GNSS stations in the CORS network of Hong Kong are selected in
162 the experiment. Those two periods are selected because they correspond to a heavy rainfall event
163 and a no-rainfall event, respectively according to hourly rainfall data from 45 rain gauges evenly
164 distributed across this area (Figure 1). There is a radiosonde station located in this area where the
165 radiosonde balloon is launched twice daily at UTC 00:00 and 12:00, respectively. The 20-years of
166 radiosonde data from 1998 to 2017 are used to establish the regional T_m model in this study. In
167 addition, the surface temperature and relative humidity are also selected to analyse their changes
168 during those two periods. To explain the variations of surface temperature and relative humidity,
169 the solar radiation data are also used in this study, which is derived from the CRU-NCEP Ver. 7
170 dataset. This dataset is a combination product of the CRU TS3.2 climate dataset and the NCEP
171 reanalysis data. The temporal-spatial resolution of the solar radiation dataset are four times daily
172 (UTC 00:00, 06:00, 12:00 and 18:00) and $0.5^\circ \times 0.5^\circ$, respectively.



173 GNSS observations are processed using Precise Point Positioning (PPP) data processing software
174 and the accuracy of the estimated ZTD parameters has been proved with the values of 7.2 mm and
175 8.1 mm when compared to the GAMIT (v10.5) and Bernese (v5.2) software, respectively (Zhao et
176 al., 2018a). The sampling rate of the estimated ZTD is 30 s and the data processing strategy has
177 been presented previously (Zhao et al., 2018d). In addition, the gradient parameters in south-north
178 and east-west directions are also estimated at intervals of 2 h. The corresponding meteorological
179 parameters, such as the surface pressure and temperature, are also obtained at the selected GNSS
180 stations. Therefore, the precise ZHD can be calculated by the empirical model using the observed
181 surface pressure. The conversion factor, as described in Eq. (1), is also obtained, in which T_m is
182 calculated based on the established T_m model which will be introduced in the following section.
183 Finally, the PWV time series, as well as the SWVs for the 13 selected GNSS stations, can be
184 obtained. Five of the 45 rain gauges (R21, TMS, PEN, SSP, and KSC) are selected to analyse the
185 variations in atmospheric water vapour during different weather conditions (Figure 1).



186
187 Figure 1. Geographic distribution of selected GNSS and radiosonde stations as well as the rain
188 gauges used in the experiment
189

190 3.2 Establishment of the regional T_m model

191 Due to the layered parameters such as water vapour pressure, temperature, *etc.* generally cannot be
192 obtained for the location of GNSS stations, the T_m values of those stations are calculated based on
193 the empirical model in this experiment. It has been proved that T_m is highly correlated with the

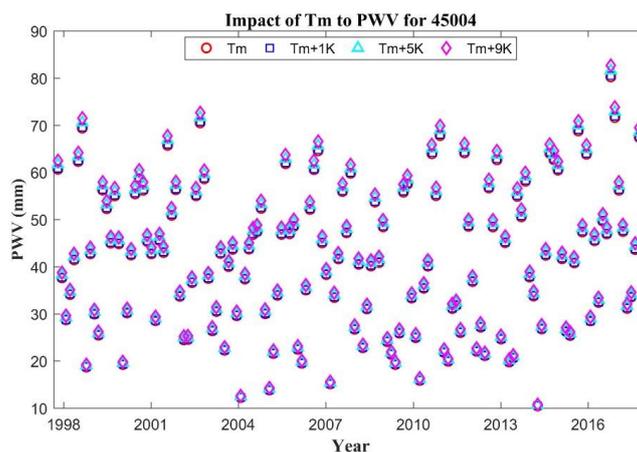


194 variations of temperature, pressure, and the seasons (Bevis et al., 1992; Yao et al., 2012; Yao et al.,
195 2014, 2015; Liu et al., 2018). Therefore, a regional T_m model which includes as parameters:
196 temperature, surface pressure, and seasonal variation, is established and expressed as follows:

$$197 \quad T_m = T_{m0} + a * T_s + b * P_s + c * \cos\left(2\pi \frac{doy}{365.25}\right) + d * \sin\left(2\pi \frac{doy}{365.25}\right) \quad (6)$$
$$\quad \quad \quad + e * \cos\left(4\pi \frac{doy}{365.25}\right) + f * \sin\left(4\pi \frac{doy}{365.25}\right)$$

198 Where T_{m0} , T_s , and P_s represent the initial value T_m , surface temperature, and surface pressure,
199 respectively; doy refers to the day of year; a and b are coefficients of T_s and P_s ,
200 respectively, while c to f refer to the coefficients of the seasonal correction function. In our
201 study, the coefficients in Eq. (6) were obtained by the least squares regression method using
202 20-year radiosonde data series for 45004 while the values of a to f are 129.1225, 0.5370,
203 -0.0023, 0.358, 0.813, -0.178, and 0.255, respectively.

204 The performance of the established T_m model is analysed and compared with the empirical
205 formula proposed by Bevis et al. (1994). Statistical result of 20-years of radiosonde data reveals
206 that the standard deviation and bias for the established T_m model and the empirical formula
207 proposed by Bevis et al. (1994) are 2.04/0.0009 K and 3.41/2.53 K, respectively, which indicates
208 that the established regional T_m model is superior to the empirical formula. The further to analyse
209 the impact of T_m model error on the calculated PWV, a comparison experiment is carried out for
210 radiosonde station 45004 with a variation in T_m of 1 K, 3 K, 5 K, 7 K, and 9 K, respectively and
211 compared with the actual PWV values. Figure 2 shows the impact of T_m error on PWV for
212 radiosonde station 45004 with a change in T_m of 1 K, 5 K, and 9 K, respectively. It can be clearly
213 seen from Figure 2 that the impact of T_m model error on PWV is negligible. Statistical analysis
214 shows that the PWV errors induced by the change in T_m of 1 K, 3 K, 5 K, 7 K, and 9 K are 0.15
215 mm, 0.45 mm, 0.75 mm, 1.04 mm, and 1.34 mm, respectively under the condition of $PWV > 0$
216 mm, while the values are 0.18 mm, 0.54 mm, 0.91 mm, 1.27 mm, and 1.63 mm, respectively when
217 $PWV > 40$ mm. Therefore, the PWV errors caused by the established T_m model in this study are
218 less than 0.4 mm and 0.5 mm when $PWV > 0$ mm and $PWV > 40$ mm, respectively. Such result is
219 deemed acceptable for the analysis of PWV variations with rainfall events (Akilan et al., 2015).



220

221 Figure 2. Impact of Tm to PWV for radiosonde station (45004) with a change in Tm by 1 K, 5 K
222 and 9 K, respectively over the period of 1998 to 2017

223

224 4 Signature of 2-d/4-d variations in atmospheric water vapour during rainfall

225 According to the recordings of 45 rain gauges derived from the Hong Kong Observatory, it is
226 continuous rains in Hong Kong for the period of 19 to 27, July 2015 with the largest rainfall more
227 than 300 mm. The weather conditions are cloudy and sunny without rainfall happened for the
228 period of 1 to 8, August 2015. Therefore, those two periods are selected in this paper to investigate
229 the variation characteristics of atmospheric water vapor.

230 4.1 Cases of 2-d PWV time series change

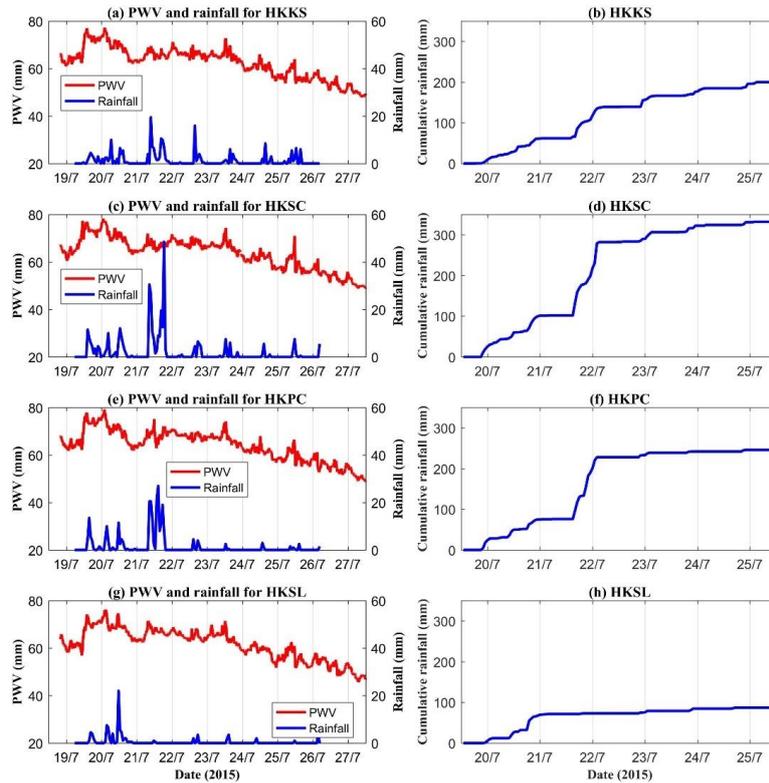
231 To capture the signature of PWV time series change in different weather conditions, the
232 comparison between the 5-minute GNSS-derived PWV and hourly rainfall are performed for the
233 periods of 19 to 27, July 2015 and 1 to 8, August 2015, respectively. Four GNSS stations (HKKS,
234 HKSC, HKPC, and HKSL) and the surrounding rainfall gauges (HSC, SSP, PEN, and R21) are
235 selected for this experiment.

236 Figure 3 shows the variations of 5-minute PWV time series data with hourly rainfall as well as the
237 cumulative rainfall at those four stations for the period of 19 to 27, July 2015 with its frequent
238 rainfall events. It can be seen, from Figure 3, that the PWV time series show an increasing trend
239 before the arrival of rainfall and reaches a relatively large value during rainfall, PWV then returns
240 to its average value after rainfall. Additionally, the PWV time series data present a downward



241 trend at four stations during this period. The cumulative rainfall first increased at about UTC 11:00,
242 20 July, 2015 with different levels reached and the event terminated at UTC 12:00, 23 July, 2015.
243 The largest cumulative rainfall reached about 250 mm while the minimum recorded rainfall was
244 about 100 mm across the four selected gauge stations. The PWV time series is also analysed at
245 those four stations for the period from 1 to 8, August, 2015 in which no rainfall was recorded
246 (Figure 4). Figure 4 shows the 5-minute PWV time series changes from which it can be found that
247 PWV does not show any continuous increasing trend when there is no rainfall, but the range of
248 PWV variation is relatively large (from about 35 mm to greater than 55 mm). Comparing the
249 PWV time series in Figures 3 and 4, it also can be observed that the PWV values during rainfall
250 are much larger than that of no rainfall time.

251 In addition, 5-minute surface temperature and relative humidity data are also analysed during
252 those two periods. The first and second columns of Figure 5 show the changes in temperature and
253 relative humidity for the period 19 to 27, July, 2015. It also can be seen that the temperature and
254 relative humidity do not show any trend during heavy rainfall but show a tendency to run counter
255 to one another on 19, 26, and 27, July. one explanation is that heavy rainfall breaks the trend in
256 temperature and relative humidity for the period from 20 to 25, July, 2015. The third column of
257 Figure 5 shows the changes in solar radiation for this period, from which it can be observed that
258 the solar radiation undergoes a day periodic change. To verify this explanation, the variations of
259 temperature and relative humidity, as well as those in solar radiation, are also presented at those
260 four stations for period without rainfall (Figure 6): temperature and solar radiation show a similar
261 trend while relative humidity presents the opposite trend. Additionally, it can be observed from the
262 first and third columns of Figures 6 that the maximum values of solar radiation and temperature
263 occurred at UTC 4:00 (local time 12:00) while the minimum value of relative humidity also
264 occurred at that time. The phenomenon found in Figure 6 further confirmed the explanation
265 presented above. In addition, the values of solar radiation are more fluctuated at the four stations
266 during rainfall when compared to that without rain (Figures 4 and 6): a possible reason for this is
267 that the part of solar radiation is decreased by cloud cover during heavy rain.



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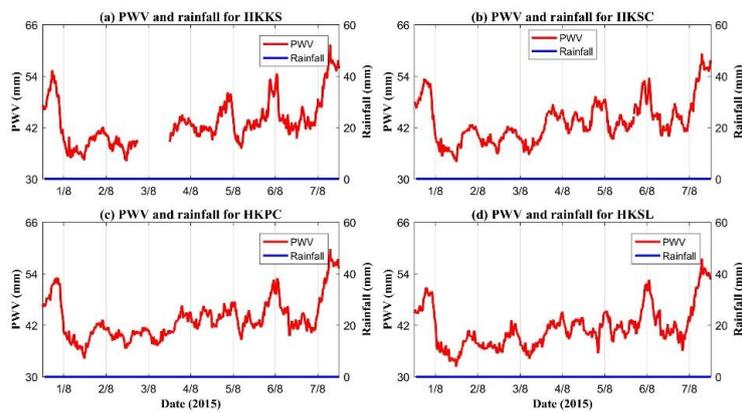
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Figure 3. Variations of 5-minutely PWV time series with hourly rainfall and the cumulative rainfall for HKKS, HKSC, HKPC and HKSL stations over the period of 19 to 27, July 2015, the first column represents the variations of PWV and rainfall and the second column refers to the cumulative rainfall



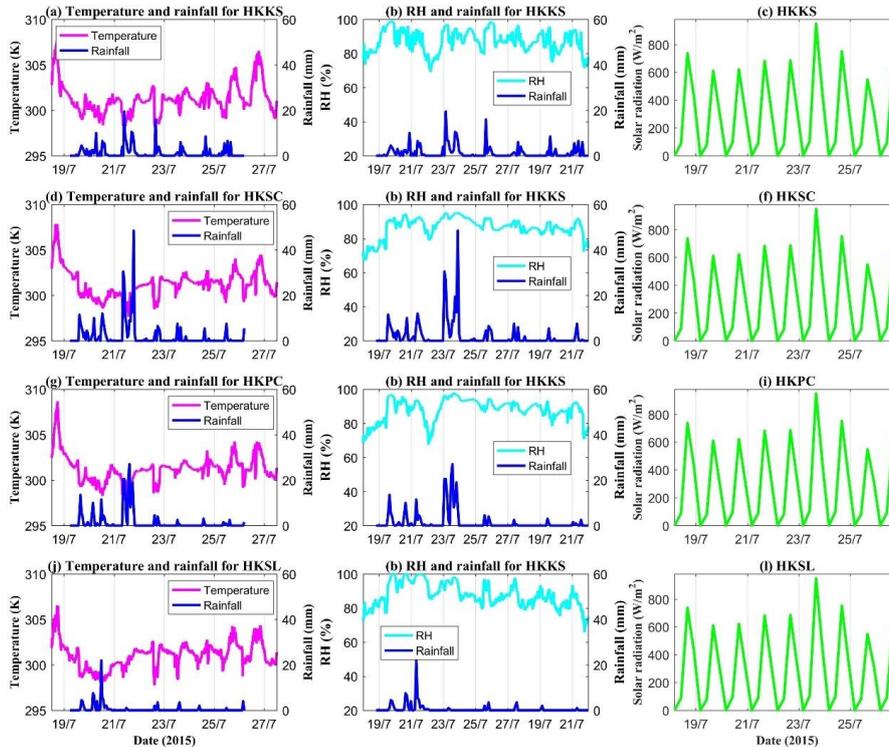
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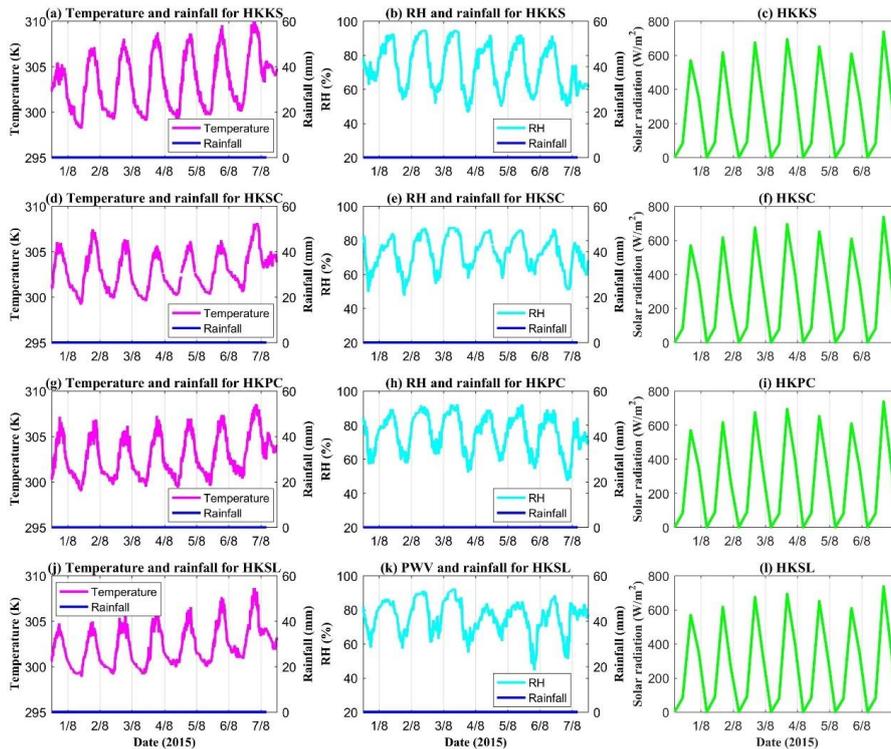
Figure 4. Variations of 5-minutely PWV time series with hourly rainfall at (a) HKKS, (b) HKSC,



275 (c) HKPC and (d) HKSL stations over the period of 1 to 7, August 2015
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277
 278
 279 Figure 5. Changes of temperature, relative humidity with rainfall as well as the solar radiation at
 280 HKKS, HKSC, HKPC and HKSL stations over the period of 19 to 27, July 2015, the first column
 281 represents the variations of temperature and rainfall, the second column refers to the variations of
 282 RH and rainfall and the third column refers to the solar radiation
 283



284

285 Figure 6. Changes of temperature, relative humidity with rainfall as well as the solar radiation at
 286 HKKS, HKSC, HKPC and HKSL stations over the period of 1 to 7, August 2015, the first column
 287 represents the variations of temperature and rainfall, the second column refers to the variations of
 288 RH and rainfall and the third column refers to the solar radiation

289

290

291 4.2 Cases of water vapour profile variation during heavy rainfall

292 The variations in 4-d atmospheric water vapour are also analysed during heavy rainfall. In this
 293 section, the tomographic technique is introduced and the research area is discretised. There are 7
 294 and 8 grids in longitudinal and latitudinal directions, respectively and 29 layers in vertical
 295 direction. Therefore, there are total $7 \times 8 \times 29$ voxels. The horizontal steps are 0.05° and 0.06° in
 296 longitudinal and latitudinal directions, respectively while the inhomogeneous vertical step is
 297 selected based on the water vapour distribution at different altitudes (Yao and Zhao, 2017) with
 298 resolutions of $0.2 \text{ km} \times 10$, $0.3 \text{ km} \times 8$, $0.4 \text{ km} \times 6$, $0.6 \text{ km} \times 4$, and $0.8 \text{ km} \times 1$, respectively. A

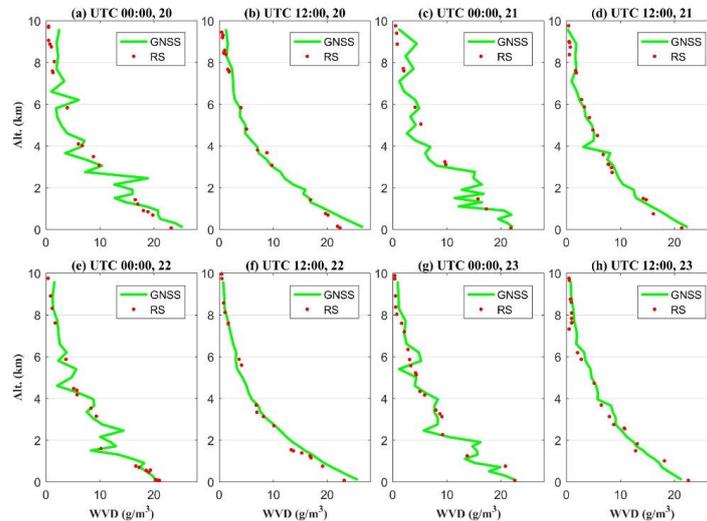


299 comparison of water vapour density profiles derived from tomographic result and radiosonde data
300 at the location of radiosonde station 45004 is first presented (Figure 7) to validate the performance
301 of the GNSS tomographic technique. It can be seen from the Figure 7 that the profiles derived
302 from tomographic results were consistent with the observed radiosonde data at most altitudes,
303 which manifests the ability of the GNSS tomographic technique to reflect variations in water
304 vapour content during rainfall. The detailed information about the accuracy of tomographic result
305 has been presented in Zhao et al. (2018d). In addition, it also can be observed that more
306 sophisticated water vapour variations detected vertically (with 29 layers) can be provided by the
307 GNSS tomographic technique than by radiosonde data.

308 Two heavy rainfall periods are selected in this experiment: the first at UTC 18 to 22, 21 July 2015
309 and three rain gauges are used to analyse the variations in water vapour profiles. The hourly
310 rainfall for those three rain gauges is presented in Table 1 while the water vapour profile variations
311 over time for SPP, PEN, and TKL are shown in Figures 8 to 10. From those three figures it can be
312 observed that atmospheric water vapour profile undergoes vertical movement about 1-2 hours
313 before the arrival of heavy rain, which is reflected by the fluctuating water vapour density at
314 different altitudes. For the SPP rain gauge, it can be seen that the water vapour content in the
315 lower atmosphere, from an altitude of about 1.8-2.5 km to 3.5 km while the water vapour content
316 decreases from 4-5 km to 3.5 km. It can also be observed from PEN and TKL rain gauges that an
317 upward and downward movement happened in the atmospheric water vapour profile in the lower,
318 and upper atmosphere, respectively: this results in a large increase in atmosphere water vapour at
319 altitudes of about 2.3 km and 1.6 km, respectively (especially at PEN). The upward and downward
320 motions of atmospheric water vapor in the lower and upper atmosphere are expected to the
321 occurrence of the strong convective weather. In addition, it was found that the variations of water
322 vapour profiles in vertical direction at station TKL are weaker than that from stations PEN and
323 SPP. A possible explanation is that the rainfall was 30.5 mm and 20.5 mm for PEN and SPP at
324 UTC 20, 21 July 2015 while the value is only 1 mm at station TKL at UTC 21, 21 July 2015
325 (Table 1). The above phenomenon indicates that the significant vertical motion of water vapour
326 profile was possibly induced by heavy rainfall. The variations in water vapour profiles during
327 rainfall reveal that the significant vertical motion of water vapour occurred before the onset of
328 rainfall while the water vapour profiles were relatively stable during rainfall events.



329 In addition, the time series of water vapour density profiles, at a temporal resolution of 1 minute,
 330 for the three rain gauges are also presented in Figure 11. From which it can be seen that the
 331 vertical water vapour density profile undergoes a significant vertical motion about 1-2 hours
 332 before the arrival of rain (black dotted rectangles, Figure 11) while the profiles are relatively
 333 stable during rain. By comparing the Figures 8-10, it also can be found that the vertical variations
 334 of water vapour density profiles at SPP and PEN stations 1 hour before rainfall are more active
 335 than that at TKL station: this can be explained by considering that the continued heavy rainfall
 336 happened at SPP and PEN stations while the TKL had little rainfall (Table 1), therefore, the
 337 continuing water vapour transportation in the vertical direction existed in the lower atmosphere at
 338 stations SPP and PEN.



339
 340 Figure 7. Distribution of water vapor density (WVD) profiles at UTC 00:00 and 12:00,
 341 respectively derived from the GNSS tomographic result (green curve) and the radiosonde data of
 342 the observed height (red hot) for the location of radiosonde station (45004) over the period of 20
 343 to 23, July 2015

344
 345 Table 1 Hourly Rainfall information of the selected four rain gauges over period of UTC 19 to 23,
 346

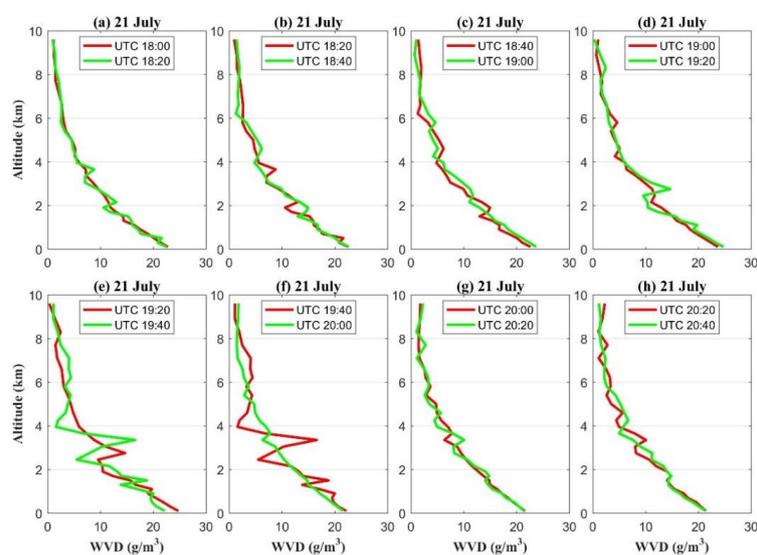
21 July 2015 (Unit: mm)

Station	SPP	PEN	TKL
Date			



19, 21 July	0	0	0
20, 21 July	30.5	20.5	0
21, 21 July	26.5	20.5	1.0
22, 21 July	10.5	13.5	1.5
23, 21 July	7.5	2.5	1.5

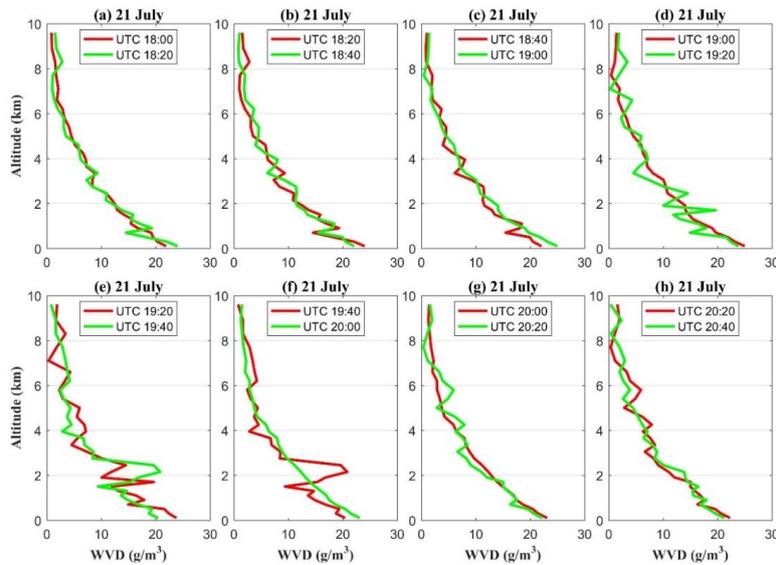
347



348

349 Figure 8. Distribution of water vapor density Profiles (WVD) derived from GNSS tomographic
 350 result with the temporal resolution of 20 minutes for the location of SPP rain gauge over the
 351 period of UTC 18:00 to 20:40, 21 July 2015

352



353

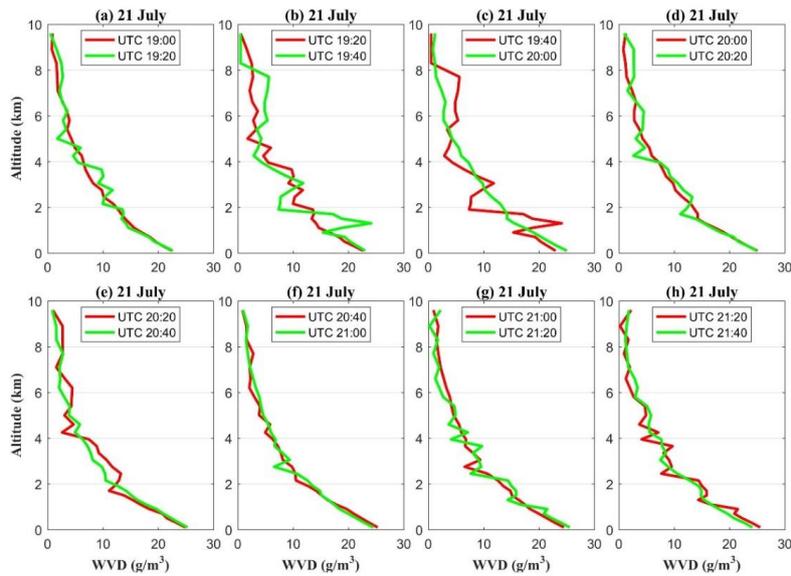
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Figure 9. Distribution of water vapor density Profiles (WVD) derived from GNSS tomographic result with the temporal resolution of 20 minutes for the location of PEN rain gauge over the period of UTC 18:00 to 20:40, 21 July 2015



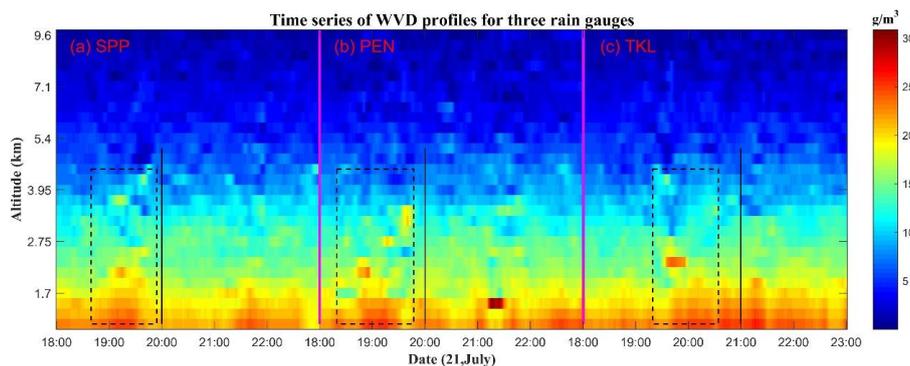
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Figure 10. Distribution of water vapor density Profiles (WVD) derived from GNSS tomographic result with the temporal resolution of 20 minutes for the location of TKL rain gauge over the period of UTC 19:00 to 21:40, 21 July 2015



362

363 Figure 11. Time series of water vapor density (WVD) profiles derived from GNSS tomographic
364 result with the temporal resolution of 1 minute for the locations of three rain gauges over the
365 period of UTC 18:00 to 23:00, 21 July 2015, where the rainfall happened at UTC 20:00 21 July
366 for (a) SPP and (b) PEN rain gauges while the rainfall occurred at UTC 21:00 21 July for (c) TKL
367 rain gauge. The WVD profiles with drastic vertical motion are marked by the black dotted
368 rectangles for three rain gauges while the locations of black solid lines are the starting time of
369 rainfall

370

371 To verify the phenomenon observed above, another period (UTC 0 to 4, 23 July 2015) at PEN and
372 TMS rain gauges is selected while the hourly rainfall information is presented in Table 2. Figures
373 12 and 13 both reflect that the change in water vapour profiles at PEN and TMS stations are
374 similar to that of above conditions. The water vapour content above PEN and TMS is increased at
375 altitudes of 2.5 km and 3.2 km, respectively, some 1-2 hours before onset of rainfall and returns to
376 its average value at the moment that the rainfall is about to begin. One possible explanation for
377 this is that: before onset of rainfall, the atmospheric water vapour was conditionally unstable with
378 intense vertical movement as proved by Brenot et al., (2006). The ascending motion of water
379 vapour in the lower atmosphere and the descending motion of water vapour in the upper
380 atmosphere significantly increases the water vapour content at a certain height where
381 hydrometeors are formed. The hydrometeors consist of liquid water and icy hydrometeors,
382 formation of which is random in time and space. Due to the delays to satellite signals induced by
383 liquid water and icy species generally being much smaller than the water vapour species-induced
384 delays, these are unavailable in the case of GNSS observations, therefore, GNSS tomography



385 cannot reflect the distribution of hydrometeors and the tomographic profiles show a
 386 returning-to-the-mean trend after the formation of hydrometeors. These newly-generated
 387 hydrometeors particles form raindrops with a continual accretion thereof. When the atmosphere is
 388 unable to support the weight of the formed raindrop, the drop falls as rain. The formation of
 389 hydrometeors particles and raindrops require some time, hence the intense vertical movement of
 390 atmospheric water vapour before onset of rainfall. The time taken to generate hydrometeors and
 391 raindrops provides the possibility of now-casting rainfall based on the GNSS technique.

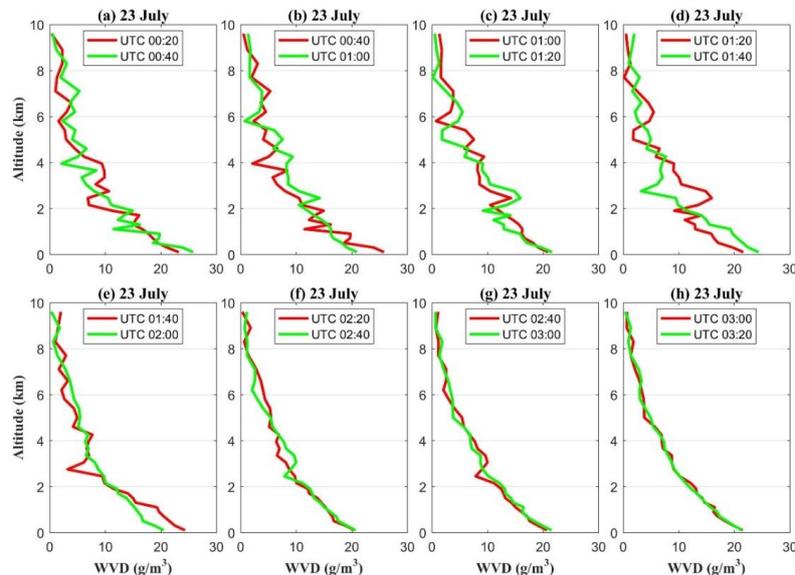
392

393 Table 2. Hourly Rainfall for the selected four rain gauges over the period UTC 1 to 5, 23 July

394

2015 (Unit: mm)		
Station	PEN	TMS
Date		
0, 23 July	0	0
1, 23 July	0	0
2, 23 July	4.5	16.5
3, 23 July	0.5	4.5
4, 23 July	0	0.5

395



396

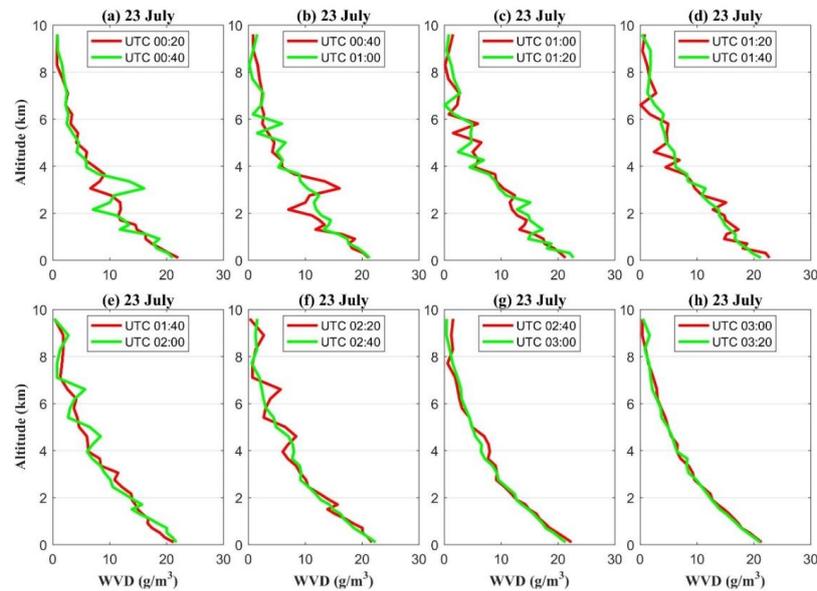
397 Figure 12. Distribution of water vapor density (WVD) profiles from GNSS tomographic result
 398 with the temporal resolution of 20 minutes for the location of PEN rain gauge over the period of



399

UTC 00:20 to 03:20, 23 July 2015

400



401

402 Figure 13. Distribution of water vapor density (WVD) profiles from GNSS tomographic result

403 with the temporal resolution of 20 minutes for the location of TMS rain gauge over the period of

404 UTC 00:20 to 03:20, 23 July 2015

405

406 The 4-d distribution of atmospheric water vapour for the period UTC 18:00 to 20:20, 21 July 2015

407 is presented with a spatio-temporal resolution of 20 minutes and 20 layers to an altitude of 5 km,

408 respectively (Figure 14). According to the hourly rainfall recordings at 45 rain gauges in this area,

409 most parts of the experimental area suffered heavy rainfall at UTC 20:00, 21 July 2015 that lasted

410 for several hours. It can be found, from Figure 14, that the significant vertical motion of water

411 vapour observed over the period from UTC 18:00 to 19:40 returns to its relatively stable condition

412 at UTC 20:00 but with a lower water vapour content in most layers. The main reason for this may

413 be water vapour transfer to the liquid water particles and icy hydrometeors, which have little

414 impact on the delay of satellite signals and cannot be observed by the GNSS technique. For the

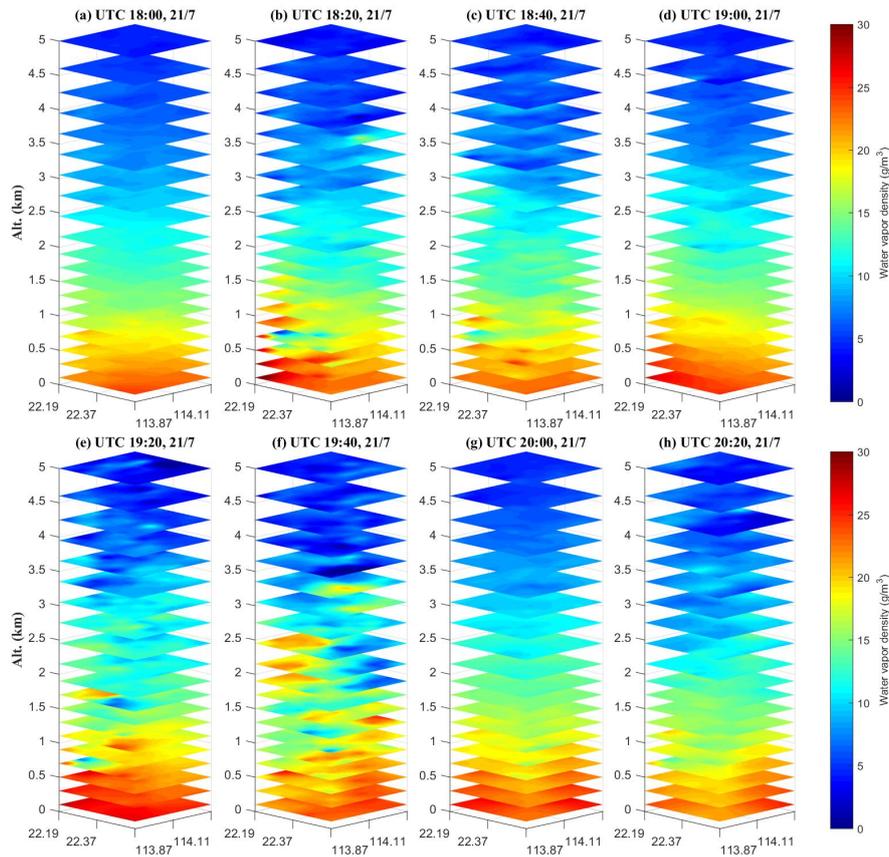
415 period of heavy rainfall that occurred after UTC 20:00, the atmospheric water vapour profiles

416 were relatively stable with slight vertical variation in water vapour content. In addition, it can be

417 concluded that the place at which hydrometeors were generated in the lower atmosphere is



418 possibly where rainfall occurred. Therefore, where heavy rainfall occurred is possibly predictable
419 before the onset of rainfall according to the 4-d atmospheric water vapour variations at different
420 altitudes derived from GNSS tomography. It also can be found that there is the horizontal motion
421 of atmospheric water vapor as well in different layers, especially at the bottom layers. This is
422 because the happening of rainfall requires the enough water vapor supplement, the horizontal
423 motion of water vapor at the bottom layers implies the continuous water vapor transportation.



424
425 Figure 14. Three-dimensional distribution of atmospheric water vapor density derived from GNSS
426 tomographic result with the temporal resolution of 20 minutes over the period of UTC 18:00 to
427 20:20, 21 July 2015 with 20 layers from the ground to 5km

428

429

430 5 Conclusion



431 GNSS sensing water vapour is an effective, practical technique, which able to the reflect 2-d and
432 4-d atmospheric water vapour variations during the formation and lifecycle of heavy rainfall. 2-d
433 PWV time series data derived from GNSS observations are first compared with hourly rainfall
434 measurements, which reveals the continuous increasing trend in PWV before the onset of rainfall
435 and returns to its average value after rainfall. In addition, it is also found that the variations of
436 surface temperature and relative humidity have day-periodicity and are mainly caused by the
437 variations in solar radiation during no rain periods, but their changes are disturbed by rainfall
438 during rainfall periods.

439 A 4-d water vapour reconstruction technique is performed using GNSS data to analyse the vertical
440 water vapour movement during rainfall period. It is found that significant vertical motion occurred
441 about 1-2 hours before the arrival of rainfall and this was reflected by the ascending and
442 descending motions of water vapour in the lower and upper atmosphere, respectively.
443 Hydrometeors are then formed at a certain altitude where sufficient water vapour was concentrated.
444 The formation of hydrometeors and raindrops requires some time, which makes it possible for the
445 forecasting of now-casting rainfall. At the moment of onset of rainfall, the water vapour profiles
446 return to their average values at different altitudes and show a relative stable condition but with a
447 decreasing trend in the water vapour content in the lower atmosphere. In addition, the place where
448 the rainfall is most possible happened may be forecasted by locating out the location of the point
449 of decreasing water vapour content in the lower atmosphere. These results revealed that rainfall
450 had a direct relationship with atmospheric water vapour content as well as the vertical variations
451 of water vapour density profiles, which further manifested the significant potential of the GNSS
452 technique for monitoring and forecasting during the lifecycle of rainfall event.

453

454

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460

461 **References:**

- 462 Akilan A.; Azeez K K A.; Balaji S, et al. GPS derived Zenith Total Delay (ZTD) observed at tropical locations in
463 South India during atmospheric storms and depressions. *Journal of Atmospheric and Solar-Terrestrial Physics*.
464 2015, 125, 1-7.
- 465 Alshawaf, F. Constructing water vapor maps by fusing InSAR, GNSS and WRF data. Karlsruhe, Karlsruher
466 Institut für Technologie (KIT), Dissertation. 2013.
- 467 Askne J.; Nordius H. Estimation of tropospheric delay for microwaves from surface weather data[J]. *Radio*
468 *Science*. 1987, 22(03), 379-386.
- 469 Bai, Z. “Near-real-time GPS sensing of atmospheric water vapour,” Ph.D. dissertation, Queensland Univ. of
470 Technol., Brisbane, Queensland, 2004
- 471 Barindelli, S.; Realini, E.; Venuti, G.; Fermi, A & Gatti, A. Detection of water vapor time variations associated
472 with heavy rain in northern Italy by geodetic and low-cost GNSS receivers. *Earth Planets & Space*, 2018, 70(1),
473 28.
- 474 Bender M.; Raabe A. Preconditions to ground based GPS water vapour tomography. *Annales geophysicae*. 2007,
475 25(8), 1727-1734.
- 476 Bender M.; Stosius R.; Zus F.; et al. GNSS water vapour tomography—Expected improvements by combining GPS,
477 GLONASS and Galileo observations. *Advances in Space Research*. 2011, 47(5), 886-897.
- 478 Bender M.; Dick G.; Ge M.; et al. Development of a GNSS water vapour tomography system using algebraic
479 reconstruction techniques. *Advances in Space Research*, 2011, 47(10), 1704-1720.
- 480 Benevides P.; Catalao J.; Miranda P M A. On the inclusion of GPS precipitable water vapour in the nowcasting of
481 rainfall. *Natural Hazards and Earth System Sciences*. 2015b, 15(12), 2605-2616.
- 482 Benevides P.; Nico G.; Catalao J.; et al. Merging SAR interferometry and GPS tomography for high-resolution
483 mapping of 3D tropospheric water vapour. *Geoscience and Remote Sensing Symposium (IGARSS), 2015 IEEE*
484 *International*. IEEE, 2015a, 3607-3610.
- 485 Bennitt G V.; Jupp A. Operational assimilation of GPS zenith total delay observations into the Met Office
486 numerical weather prediction models. *Monthly Weather Review*, 2012, 140(8), 2706-2719.
- 487 Bevis M.; Businger S.; Herring T A.; et al. GPS meteorology: Remote sensing of atmospheric water vapor using
488 the Global Positioning System. *Journal of Geophysical Research: Atmospheres*. 1992, 97(D14), 15787-15801.
- 489 Bevis M.; Businger S.; Chiswell S.; et al. GPS meteorology: Mapping zenith wet delays onto precipitable water[J].
490 *Journal of applied meteorology*, 1994, 33(3), 379-386.
- 491 Bi Y.; Mao J.; Li C. Preliminary results of 4-D water vapor tomography in the troposphere using GPS. *Advances in*
492 *atmospheric sciences*, 2006, 23(4), 551-560.
- 493 Braun, John Joseph. Remote sensing of atmospheric water vapor with the global positioning system. *Geophysical*
494 *Research Letters*, 2004, 20(23), 2631-2634.
- 495 Braun J.; Rocken C.; Liljegren J. Comparisons of line-of-sight water vapor observations using the global
496 positioning system and a pointing microwave radiometer. *Journal of Atmospheric and Oceanic Technology*, 2003,
497 20(5), 606-612.
- 498 Brenot H.; Neméghaire J.; Delobbe L.; et al. Preliminary signs of the initiation of deep convection by GNSS.
499 *Atmospheric Chemistry and Physics*, 2013, 13(11), 5425-5449.
- 500 Brenot, H., Ducrocq, V., Walpersdorf, A., Champollion, C., & Caumont, O. (2006). GPS zenith delay sensitivity
501 evaluated from high - resolution numerical weather prediction simulations of the 8 - 9 September 2002 flash flood
502 over southeastern France. *Journal of Geophysical Research: Atmospheres*, 111(D15).
- 503 Chen B, Liu Z. Voxel-optimized regional water vapor tomography and comparison with radiosonde and numerical
504 weather model. *Journal of geodesy*, 2014, 88(7), 691-703.
- 505 Dach R. and Walser P. “Bernese GNSS software version 5.2,” Astronomical Inst., Univ. of Bern, Bern, 2013.



- 506 De Haan S. Assimilation of GNSS ZTD and radar radial velocity for the benefit of very-short-range regional
507 weather forecasts. *Quarterly Journal of the Royal Meteorological Society*. 2013, 139(677), 2097-2107.
- 508 Flores A.; Ruffini G.; Rius A. 4D tropospheric tomography using GPS slant wet delays//*Annales Geophysicae*.
509 *Springer-Verlag*, 2000, 18(2), 223-234.
- 510 Guerova G.; Jones J.; Dousa J.; et al. Review of the state of the art and future prospects of the ground-based GNSS
511 meteorology in Europe. *Atmospheric Measurement Techniques*, 2016, 9(11), 1-34.
- 512 Heublein M.; Zhu X X.; Alshawaf F.; et al. Compressive sensing for neutrospheric water vapor tomography using
513 GNSS and InSAR Observations[C]//*Geoscience and Remote Sensing Symposium (IGARSS), 2015 IEEE*
514 *International*. IEEE, 2015, 5268-5271.
- 515 Hirahara K. Local GPS tropospheric tomography. *Earth, planets and space*, 2000, 52(11), 935-939.
- 516 JMA (2013) Outline of the operational numerical weather prediction at the Japan Meteorological Agency.
517 Appendix to WMO technical progress report on the global data-processing and forecasting system (GDPFS) and
518 numerical weather prediction (NWP) research.
519 <http://www.jma.go.jp/jma/jma-eng/jma-center/nwp/outline2013-nwp/index.htm>. Accessed 30 Aug 2017
- 520 Liu J.; Yao Y.; Sang J. A new weighted mean temperature model in China. *Advances in Space Research*, 2018,
521 61(1), 402-412.
- 522 Liu Z.; Wong M S.; Nichol J.; et al. A multi-sensor study of water vapour from radiosonde, MODIS and
523 AERONET: a case study of Hong Kong. *International Journal of Climatology*, 2013, 33(1), 109-120.
- 524 Perler D.; Geiger A.; Hurter F. 4D GPS water vapor tomography: new parameterized approaches. *Journal of*
525 *Geodesy*, 2011, 85(8), 539-550.
- 526 Rius A.; Ruffini G.; and Cucurull L.; “Improving the vertical resolution of ionospheric tomography with GPS
527 occultations,” *Geophys. Res. Lett.* 1997, 24(18), 291–2294.
- 528 Rohm W, Bosy J. Local tomography troposphere model over mountains area. *Atmospheric Research*, 2009, 93(4),
529 777-783.
- 530 Rohm W.; Bosy J. The verification of GNSS tropospheric tomography model in a mountainous area . *Advances in*
531 *Space Research*, 2011, 47(10), 1721-1730.
- 532 Saastamoinen, J. Atmospheric correction for the troposphere and stratosphere in radio ranging satellites. *The use of*
533 *artificial satellites for geodesy*, 1972, 247-251.
- 534 Saito K.; Shoji Y.; Origuchi S.; et al. GPS PWV Assimilation with the JMA Nonhydrostatic 4DVAR and Cloud
535 Resolving Ensemble Forecast for the 2008 August Tokyo Metropolitan Area Local Heavy Rainfalls//*Data*
536 *Assimilation for Atmospheric, Oceanic and Hydrologic Applications (Vol. III)*. Springer, Cham, 2017, 383-404.
- 537 Seko H.; Shimada S.; Nakamura H.; et al. Three-dimensional distribution of water vapor estimated from
538 tropospheric delay of GPS data in a mesoscale precipitation system of the Baiu front [J]. *Earth, planets and space*,
539 2000, 52(11), 927-933.
- 540 Skone S.; Hoyle V. Troposphere Modeling in a Regional GPS Network. *Positioning*, 2005, 4(1&2), 230-239.
- 541 Troller M.; Geiger A.; Brockmann E.; et al. Determination of the spatial and temporal variation of tropospheric
542 water vapour using CGPS networks. *Geophysical Journal International*, 2006, 167(2), 509-520.
- 543 Troller M.; Burki B.; Cocard M.; Geiger A.; et al. 3-D refractivity field from GPS double difference tomography.
544 *Geophysical Research. Letters*, 2002, 29, 2149–2152.
- 545 Wang, X., Wang, X., Dai, Z., et al. Tropospheric wet refractivity tomography based on the BeiDou satellite system.
546 *Advances in Atmospheric Sciences*, 2014, 31(2), 355-362.
- 547 Yao Y B.; Zhu S.; Yue S Q. A globally applicable, season-specific model for estimating the weighted mean
548 temperature of the atmosphere. *Journal of Geodesy*, 2012, 86(12), 1125-1135.
- 549 Yao Y B.; Zhao Q Z.; Zhang B. A method to improve the utilization of GNSS observation for water vapor
550 tomography. *Annales Geophysicae (09927689)*, 2016, 34(1), 143-152.



- 551 Yao Y.; Zhao Q. Maximally Using GPS Observation for Water Vapor Tomography. *IEEE Transactions on*
552 *Geoscience and Remote Sensing*, 2016, 54(12), 7185-7196.
- 553 Yao Y.; Zhao Q. A novel, optimized approach of voxel division for water vapor tomography[J]. *Meteorology and*
554 *Atmospheric Physics*, 2017, 129(1), 57-70.
- 555 Yao Y.; Shan L.; Zhao Q. Establishing a method of short-term rainfall forecasting based on GNSS-derived PWV
556 and its application. *Scientific reports*, 2017, 7(1), 12465.
- 557 Yao Y.; Zhang B.; Xu C. and Yan, F. Improved one/multi-parameter models that consider seasonal and geographic
558 variations for estimating weighted mean temperature in ground-based GPS meteorology. *Journal of Geodesy*, 2014,
559 88(3), 273-282.
- 560 Yao Y B.; Liu J H.; Zhang B.; et al. Nonlinear relationships between the surface temperature and the weighted
561 mean temperature. *Geomatics & Information Science of Wuhan University*, 2015, 40(1), 112-116.
- 562 Zhang K.; Manning T.; Wu S.; et al. Capturing the signature of severe weather events in Australia using GPS
563 measurements. *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 2015, 8(4),
564 1839-1847.
- 565 Zhao Q.; Yao Y. An improved troposphere tomographic approach considering the signals coming from the side
566 face of the tomographic area//Annales Geophysicae. *Copernicus GmbH*, 2017b, 35(1), 87-95.
- 567 Zhao Q.; Yao Y.; Cao X.; et al. Accuracy and reliability of tropospheric wet refractivity tomography with GPS,
568 BDS, and GLONASS observations. *Advances in Space Research*, 2018c. DOI:10.1016/j.asr.2018.01.021
- 569 Zhao Q.; Yao Y.; Yao W. A troposphere tomography method considering the weighting of input
570 information[C]//Annales Geophysicae. *Copernicus GmbH*, 2017a, 35(6), 1327-1340.
- 571 Zhao Q.; Yao Y.; Yao W. GPS-based PWV for precipitation forecasting and its application to a typhoon event.
572 *Journal of Atmospheric and Solar-Terrestrial Physics*, 2018a, 167, 124-133.
- 573 Zhao Q.; Yao Y.; Cao X.; et al. An Optimal Tropospheric Tomography Method Based on the Multi-GNSS
574 Observations. *Remote Sensing*, 2018d, 10(2), 234.
- 575 Zhao Q.; Yao Y.; Yao W.; et al. Real-time precise point positioning-based zenith tropospheric delay for
576 precipitation forecasting. *Scientific reports*, 2018b, 8(1), 7939-7939.