GPS water vapour tomography: preliminary results from the ESCOMPTE field experiment

C. Champollion\textsuperscript{a,}\textsuperscript{*}, F. Masson\textsuperscript{a}, M.-N. Bouin\textsuperscript{b}, A. Walpersdorf\textsuperscript{c}, E. Doerflinger\textsuperscript{a}, O. Bock\textsuperscript{d}, J. Van Baelen\textsuperscript{e}

\textsuperscript{a}Lab. Dynamique de la Lithosphère/CNRS, Montpellier, France
\textsuperscript{b}Institut Géographique National, Saint-Mandé, France
\textsuperscript{c}Lab. Géophysique Interne et Tectonophysique/CNRS, Grenoble, France
\textsuperscript{d}Service d’Aéronomie/CNRS, Université Paris VI, Paris, France
\textsuperscript{e}CNRM/GAME, Météo-France/CNRS, Toulouse, France

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Abstract

Water vapour plays a major role in atmospheric processes but remains difficult to quantify due to its high variability in time and space and the sparse set of available measurements. The GPS has proved its capacity to measure the integrated water vapour at zenith with the same accuracy as other methods. Recent studies show that it is possible to quantify the integrated water vapour in the line of sight of the GPS satellite. These observations can be used to study the 3D heterogeneity of the troposphere using tomographic techniques. We develop three-dimensional tomographic software to model the three-dimensional distribution of the tropospheric water vapour from GPS data. First, the tomographic software is validated by simulations based on the realistic ESCOMPTE GPS network configuration. Without a priori information, the absolute value of water vapour is less resolved as opposed to relative horizontal variations. During the ESCOMPTE field experiment, a dense network of 17 dual frequency GPS receivers was operated for 2 weeks within a 20×20-km area around Marseille (southern France). The network extends from sea level to the top of the Etoile chain (~700 m high). Optimal results have been obtained with time windows of 30-min intervals and input data evaluation every 15 min. The optimal grid for the ESCOMPTE geometrical configuration has a horizontal step size of 0.05°×0.05° and 500 m vertical step size. Second, we have compared the results of real data inversions with independent observations. Three inversions have been compared

\textsuperscript{*} Corresponding author. Tel.: +33 4 67 14 45 91; fax: +33 4 67 52 39 08.
E-mail address: Cedric.Champollion@dstu.univ-montp2.fr (C. Champollion).

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to three successive radiosonde launches and shown to be consistent. A good resolution compared to
the a priori information is obtained up to heights of 3000 m. A humidity spike at 4000-m altitude
remains unresolved. The reason is probably that the signal is spread homogeneously over the whole
network and that such a feature is not resolvable by tomographic techniques. The results of our pure
GPS inversion show a correlation with meteorological phenomena. Our measurements could be
related to the land–sea breeze. Undoubtedly, tomography has some interesting potential for the water
vapour cycle studies at small temporal and spatial scales.

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1. Introduction

Water vapour plays a major role in atmospheric processes but remains difficult to
quantify due to its high variability in time and space and the sparse set of available
measurements. Therefore, the role of water vapour in precipitation, energy transfer, or
radiation budget remains poorly described. The GPS has proved its capacity to measure
the integrated water vapour at zenith with the same accuracy as other methods
(radiosondes (Tregoning et al., 1998), water vapour radiometers (Rocken et al., 1995;
Doerflinger et al., 1998)). The GPS has shown to be of great interest in meteorology and in
climatology (Bevis et al., 1992; Yuan et al., 1993; Businger et al., 1996; Duan et al., 1996;
Fang et al., 1998; Tregoning et al., 1998; Ware et al., 2000; Jerret and Nash, 2001;
Gradinarsky et al., 2002) because of its (increasing) spatial density and its continuous
recording of the state of the water vapour in the atmosphere. Compared to water vapour
radiometers, GPS operates in all weather conditions without need of complicated
calibration. The assimilation of GPS measurements in numerical weather prediction
models is under investigation in several laboratories (Gutman et al., 2004) and should
improve the description of the humidity field and the precipitation forecasts. Recent
studies show that it is possible to quantify the integrated water vapour in the line of sight
of the GPS satellite (e.g., Ware et al., 1997; Alber et al., 2000; Braun et al., 2003). These
observations can be used to study the 3D heterogeneity of the troposphere using
tomographic techniques.

Tropospheric tomography consists of the retrieval of a 3D scalar field of the water
vapour from integrated measurements. Tomographic techniques have been intensively used
in recent decades in medicine to investigate the human body or in seismology to describe
the seismic velocity anomaly of the Earth’s interior. In these two domains, tomographic
inversion is well constrained by a sufficient amount of data which can be accumulated
because the structures investigated do not change in time. The tomography of tropospheric
water vapour is less well constrained due to the limited number of simultaneously visible
GPS satellites and the great temporal variability of the water vapour. The first tropospheric
tomography studies of water vapour have been performed by Flores et al. (2000) and Flores
(2000) on the Kilauea volcano (Hawaii) and by Gradinarsky (2002) in the Göteborg region
(Sweden). These experiments have been able to prove the feasibility of GPS tropospheric
tomography. However, the data sets are limited by the number of GPS stations or by the
lack of other meteorological data. The dedicated network for water vapour tomography was deployed during the ESCOMPTE field experiment conducted in Marseille (southern France) in June and July 2001. A dense GPS network was installed during the extensive campaign of pollution and atmospheric measurements ESCOMPTE (more information available on the website: http://medias.obs-mip.fr/escompte). This particular setting provides a good opportunity for the validation of the tomographic solution, for its interpretation, and for meteorological valorization.

Considering the dispersive character of the ionosphere for the GPS frequencies, ionospheric effects are minimised using a fitted linear combination of the two GPS frequencies (Brunner and Gu, 1991). Conversely, the tropospheric effects are not frequency-dependent below 15 GHz. The main effect of the troposphere on GPS positioning is an extra delay of the radio signal emitted by the GPS satellites (Davis et al., 1985). This delay is time varying due to the variable pressure, temperature, and water vapour content of the atmosphere and cannot be modelled or predicted with sufficient precision for high-precision positioning, especially in near real time. This delay is nowadays the major error source for precise positioning. To model out the perturbation, a set of tropospheric parameters is estimated during the GPS data analysis: zenith delays and, more recently, horizontal gradients. The correlation between these delays and the state of the atmosphere makes the GPS an efficient tool for meteorological observation. In this study, we will describe the techniques for the retrieval of Zenith Total Delay (ZTD) and the conversion to Slant Integrated Water Vapour (SIWV) between each GPS receiver and each GPS satellite. The SIWV values are the input data for the tomography. We will discuss the implications of the accuracy of the SIWV signal for the interpretations of the results. We explain the theory of our tomographic inversion and its limitations based on synthetic results. Then, we describe the ESCOMPTE field experiment and the GPS data analysis. Finally, we show some preliminary results from the inversion of the ESCOMPTE GPS data and a first tentative of meteorological interpretation.

2. From the GPS signal to the tomographic modelling

Atmospheric measurements with GPS are based on the propagation delay of the signal through the real atmosphere with respect to the propagation of the same signal in a vacuum. The tropospheric delay $\Delta L$ is obtained by integrating the neutral refractivity $N(l)$ along the path $L$ between the satellite and the receiver:

$$\Delta L = \int_L 10^{-6}Ndl$$

The path $L$ is calculated from Fermat’s law of wave’s propagation. The bending of the ray is implicitly included in the mapping function described later. The refractivity can be expressed as a function of the “dry” air pressure $P_d$, the temperature $T$, and the water vapour partial pressure $P_w$:

$$N = k_1 \frac{P_d}{T} Z_d^{-1} + k_2 \frac{P_w}{T} Z_w^{-1} + k_3 \frac{P_w}{T^2} Z_w^{-1}$$

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where \( Z_d^{-1} = 1.00027 \) is the inverse of the “dry” air compressibility, \( Z_w^{-1} = 1.00201 \) is the inverse of the water vapour compressibility, and \( k_1 = 77.60 \pm 0.05 \text{ K/hPa} \), \( k_2 = 70.4 \pm 2.2 \text{ K/hPa} \), \( k_3 = (3.739 \pm 0.012).10^5 \text{ K}^2/\text{hPa} \) (Bevis et al., 1994).

In the GPS data analysis, the delay is expressed at the zenith in two parts:

\[
ZTD = \text{ZHD} + \text{ZWD}
\]

where ZHD and ZWD are the hydrostatic and wet parts of the ZTD.

From Eq. (3), we can remove the “dry” or hydrostatic delay depending only on the total pressure and the temperature. The zenith hydrostatic delay \( \text{ZHD} \) is accurately estimated from the surface pressure \( P_0 \) and the variation of the gravity field \( f \) with the latitude \( \varphi \) and the height above the geoids \( H \) in kilometres (Davis et al., 1985):

\[
\text{ZHD} = \frac{P_0 f(\varphi,H)}{m_w P_0 f(\varphi,H) + k_1 m_w m_d + k_3 T_m R^*}
\]

with \( f(\varphi,H) = 1 - 0.00265 \cos 2\varphi - 0.000285H \)

The remaining zenith “wet” delay ZWD is simply the difference between the ZTD and the ZHD. The Zenith Integrated Water Vapour IWV is nearly proportional to the ZWD, with a conversion factor \( \Pi \) (in kg/m\(^3\)) expressed after Bevis et al. (1992) as

\[
\Pi = \frac{10^6 m_w}{(k_2 - k_1 \frac{m_w}{m_d} + \frac{k_3}{T_m}) R^*}
\]

where \( T_m \) is a so-called “mean temperature,” \( m_w \) is the molar mass of water vapour, \( m_d \) is the molar mass of dry air, and \( R^* \) is the ideal gas constant. The relation between the ground temperature and the mean temperature \( T_m \) can be empirically determined from a data set of regional radio soundings with an accuracy of 2% (Bevis et al., 1992). But, for this study, we assume an accuracy of 1% because we use the data from the three radio soundings launch within the GPS network at each time of the three tomographic inversions presented, and we integrate numerically the formula

\[
T_m = \int \frac{P_w}{T} dz / \int \frac{P_w}{T^2} dz
\]

Finally, we have

\[
\text{IWV} = \Pi \times \text{ZWD}
\]

The integrated water vapour in the vertical column is retrieved from slant wet delays to all the satellites observed by a ground-based GPS station. The cone on Fig. 1 is limited by the minimum elevation of the GPS observations. Comparisons with radio soundings, Very Long Baseline Interferometry, and radiometer measurements lead to an accuracy of about 0.8 kg/m\(^2\) for GPS-derived IWV (Foelsche and Kirchengast, 2001; Niell et al., 2001).
2.1. From zenith delay to Slant Integrated Water Vapour content

The zenithal delays are mapped to the zenith to form the slant delays observed between the GPS satellites and antennas with $m_h$ and $m_w$ the corresponding “dry” and “wet” mapping functions. In this study, we use the Niell hydrostatic and wet mapping functions (Niell, 1996). Any atmospheric behaviour departing from this assumed azimuthal symmetry (e.g., lateral heterogeneity) cannot be reproduced by this horizontal layer model (Bock et al., 2001). We can improve the model by estimating gradients in the North–South and East–West direction. The gradients (Fig. 1) are fit to minimise the nonmodelled residuals as described by Davis et al. (1993) and Chen and Herring (1997), and they are expressed in millimeters of delay at 10° of elevation. We use this gradients approach to introduce tropospheric heterogeneity in the GPS solution. The gradients in the hydrostatic component can be calculated (and then removed) from surface measurements. The Slant Integrated Water Vapour (SIWV) are reconstructed from ZWD and gradients, they are the input data (or observations) for our tomographic inversion. Namely, the SIWV is now expressed as

\[
\text{SIWV} = m_w(e)\text{ZWD} + m_\Delta(e)\cot(e)\left[G_N\cos\phi + G_E\sin\phi\right]
\]

with the above notations, $e$ is the elevation angle, $G_N$, $G_E$, and $m_\Delta$ being the N–S gradient of ZTD, the E–W gradient of ZTD, and the gradient mapping function, respectively (Chen and Herring, 1997). All the atmospheric variability cannot be completely described by such a model, the postfit residuals can be added to the slant wet delays. The postfit residuals are the difference between the phase observations and the signal reconstruction at each epoch between each satellite–station pair, including the clock’s error, the multipath, the phase centre variations, and the atmospheric nonmodelled effects. These components are not easily removed, even with the stacking method used by Braun et al. (2001).
Quantifying exactly the real part of the atmospheric nonmodelled effects in these residuals is not feasible yet.

As the gradients are an average over all the visible satellites, the noise level is still lower than the postfit residuals. During the ESCOMPTE campaign, the magnitude of signal level in gradients is typically about 2 mm at the zenith, with an RMS of about 0.5 mm. The noise level in the slant delay postfit residuals is equivalent compared to the signal level in the gradients. In this study, we do not use the residuals to avoid contamination from nonatmospheric GPS errors (such as multipath and clock errors) in our observations.

An alternative method is to use directly the double difference residuals which are free from clock’s errors. But the methods to extract information from double difference residuals (Alber et al., 2000) are still discussed and may cause significant systematic errors (Elòsegui and Davis, 2003).

2.2. The tomographic inversion: theory and error covariance formulation

The tomographic inverse problem consists of retrieving the scalar three-dimensional field of water vapour from the values integrated along the ray path (Fig. 2b). Since we consider the geometrical effect of the bending of the ray as negligible for elevation angles higher than $10^\circ$ (Elgered, 1993), the problem becomes linear. The formulation in the linear discrete theory of the direct problem expresses the link between the

Fig. 2. (a) Schematic view in 2D of the discretization of the space and the assumed straight rays between ground receivers and GPS satellites. (b) $\Delta$SIWV is the path-length of one ray in one cell in three dimensions.
observations (SIWV) $Y$ and the true water vapour density $X$ (expressed in g/m$^3$) through the linear operator $M$:

$$Y = M X + \varepsilon$$

(7)

where $\varepsilon$ is the measurement error vector. The linear operator $M$ is expressed as a matrix whose coefficients represent the length of each ray in each cell ($\Delta$SIWV in Fig. 2b). To solve the inverse problem, the matrix $M$ must be inverted. As the problem is locally underdetermined (some cells have no rays within), we have to find the “generalised inverse” $M^{-\varepsilon}$ to obtain the reconstructed field $X_{\text{reco}}$ from the data $Y$:

$$X_{\text{reco}} = M^{-\varepsilon} Y$$

(8)

To minimise the ill-conditioning, we can add some constraints using a priori information ($X_{\text{ap}}$). A good a priori field can be water profiles from radio soundings or radio occultation (above 2 km). As the aim of this study is to determine the ability of the tomographic method to reconstruct the water vapour field from GPS data alone, we choose an a priori profile from a standard atmosphere for midlatitude. The optimal case would be if GPS tomography were reliable in every region in every weather condition, even if no radio soundings are available. Radio sounding profiles will be used for the validation of the results. To gain some stability in the lowest layer, we use surface measurements of water vapour to initialise the ground level of the standard atmosphere. We could also add some constraints instead of the a priori model to smooth the density field (Flores et al., 2001a,b), but the level of the constraints is difficult to set and can hardly be evaluated from atmospheric physics.

The Bayesian approach in the framework of discrete inverse theory gives us an explicit manner to incorporate a priori information (Foelsche and Kirchengast, 2001; Rogers, 2000). With the assumption of a linear forward model and of a Gaussian distribution for the a priori and measurement error, we can express the reconstructed field of water vapour $X_{\text{reco}}$ as

$$X_{\text{reco}} = X_{\text{ap}} + \left[ M^T C_y^{-1} M + C_{\text{ap}}^{-1} \right]^{-1} M^T C_y^{-1} \left( Y - MX_{\text{ap}} \right)$$

(9)

where the matrices $C_y$ and $C_{\text{ap}}$ are respectively the measurement error covariance and the a priori error covariance.

The a priori covariance $C_{\text{ap}}$ is chosen to be quite large except for the lowest layer which is partially adjusted to the ground measurement. At the surface, we use an error of 25% and vary that error up to 100% at 5000 m high based on radio sounding data sets in Europe (Gradinarsky, 2002). The correlation between two elements of the a priori model is given by a classic Gaussian law in the horizontal and an exponential law in the vertical. The exponential law reflects the high vertical variability of water vapour. We choose a small vertical and horizontal length of Gaussian correlation equal to 0.1 and 0.5 km to avoid some smoothing effects.

The measurement error covariance $C_y$ can be split into three different parts: the discretization $C_{\text{dis}}$, the observation $C_{\text{obs}}$ (ZWD and conversion error to SIWV), and the spatial and temporal variations of the humidity field $C_{\text{var}}$. The discretization error $C_{\text{dis}}$ (based on direct modelling of real data into different grids) shows a white noise level of 2% for $0.05^\circ \times 0.05^\circ \times 500$ m cells, with no bias and no correlation. The ZWD error
(typically 6 mm) is projected along the ray path with the Niell mapping function to obtain $C_{\text{obs}}$ (relative to the noise $\alpha$ in Eq. (7)). The error on the $\Pi$ coefficient is based on the estimation of the temperature $T_m$ with an error of 1%. These errors do not account for any possible spatial correlation among ZWD at different sites because of the lack of knowledge in this domain (Tregoning et al., 1998). We assume that the uncorrelated errors dominate the total measurement error as we use a network with baseline longer than 2000 km. Finally, the spatial variation within a cell $C_{\text{var}}$ and the temporal variation during the time of measurement are determined from the interpolation of IWV map for the whole GPS network.

3. Parameterization of tomographic inversion based on simulated data

To examine the effective resolution of the method, a set of synthetic tests using various models has been performed. We use the real geometric configuration of the ESCOMPTE experiment. In a first step, we test the number of rays needed to solve the inverse problem. A simple set of data, corresponding to 15 min of GPS data, is formed by the SIWV between each satellite–station pair computed at one epoch. To increase the number of data, it is possible to inverse together several sets of data (three sets of data=30 min, five sets of data=1 h). It is also possible to interpolate the data between two epochs to get one set of data each 30 s. This allows a denser spatial coverage of the studied region. The best result is obtained using SIWV each 30 s over 1 h, allowing the best coverage of the 3D model as we use the maximum of data. Unfortunately, the inversion of real data is much more complex. First, accurate SIWV are needed, avoiding the use of interpolated data. Second, the inversion of five sets of data implies a time interval too long with respect to the variability of the water vapour field in the troposphere. To obtain a compromise between the quantity of data and the variability of the water vapour field, we select a time interval of 30 min. In a second step, we test the size of the grid cells. The optimal horizontal size of the cell is equal to the mean distance between the stations ($\sim 0.05^\circ$). Vertically, the thickness of the layers has to be defined to allow the rays starting from one station to cross the neighbouring cells. Taking the station at the centre of the cell and a minimum elevation angle of $10^\circ$, this implies a minimum thickness of the first layer of about 300 m. To confirm these estimations, we have performed two inversions taking cell sizes of $0.05^\circ \times 0.05^\circ \times 500$ m and $0.025^\circ \times 0.025^\circ \times 200$ m, respectively. Clearly, the set of data does not allow resolving the second grid: this fine grid is not resolvable due to the small amount of data by cell. Only a greater number of stations decreasing the mean distance between the stations could allow the use of smaller cells. Using the $0.05^\circ \times 0.05^\circ \times 500$-m grid, only 12% of the cells are empty of ray crossings.

Hereafter, we present five synthetic tests performed using a time interval of 30 min and a grid of $0.05^\circ \times 0.05^\circ \times 500$ m. We study several simple cases to illustrate the geometrical limitations of the method. All the rays between the satellites and the ground receivers must be in the grid so we add large buffer cells around the GPS network. The synthetic data are constructed from the real position of the antennas and the satellites. The water vapour density in the atmosphere is set to zero, and the anomalies to one. As we look only at the geometrical limitations of the method, there is no noise added to the synthetic data. As the
Fig. 3. (a–d): Eight horizontal slices from 250 to 3750 m high through a synthetic model (□) and the resulting inverted model (○). Anomalies in the synthetic model = 1. Values in the circles are the values of the retrieved field. See text for comments.
Fig. 3 (continued).
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Fig. 3 (continued).
atmospheric model is completely unrealistic, we simply use a null a priori model with no particular constraints for the lowest layer. Further synthetic tests will be done in the future with more realistic models from numerical weather simulations. The results of the four tests are summarized below (see Fig. 3a to d):

(1) A cross-shaped anomaly between 500 and 1000 m: the cross is well retrieved except in its eastern part due to the lack of rays in this cell that includes the Garlaban Mountain. A small amount of smearing is observed in the first layer and leads to an underestimation of the value of the anomaly in the cross. Negative values around the cross are due to the inversion scheme which produces the minimal norm solution. The total contrast in the 750-m slice is about 1, as expected. To obtain the absolute values of the humidity field, we need a realistic a priori model.

(2) A cross-shaped anomaly between 3500 and 4000 m: once again, the cross is well retrieved, with a small amount of smearing in the layer underneath and negative values around the cross. Our data set is capable to detect and recover structures at high altitude.

(3) A homogeneous layer between 2000 and 2500 m: the anomaly is more or less homogeneously distributed throughout the whole model except in the first layer. Synthetic data computed through the model can be explained by infinity of layered models with an integrated vertical anomaly of 1. As the inversion conducts to the minimum norm model, the solution spreads out the anomaly in all the layers. Only the first layer is correct due to the variation of the altitude of the stations.

(4) A homogeneous layer between 0 and 500 m: due to the variation of the altitude of the stations, the rays from the satellites to the stations do not have the same length within the anomaly. Only an anomalous layer at the bottom of the model can explain the synthetic data. This explains why the estimated field resembles the synthetic field. In the eastern part of the model, the topography and the lack of stations explain the large difference between the estimated and synthetic fields.

To summarize, the tomographic inversion is able to restore local anomalies (i.e., anomalies which do not cover the whole studied domain) whatever the height. On the other hand, an extended anomaly cannot be restored, except if it is located at low altitude, below the highest GPS station. In the estimated fields, we have to take into account the possible existence of vertical smearing and horizontal smoothing of the anomalies. Nevertheless, the horizontal contrasts are well retrieved and can be used to detect lateral water vapour anomalies.

4. The ESCOMPTE field experiment: preliminary tomographic results

The ESCOMPTE campaign (Cros et al., 2004; http://medias.obs-mip.fr/escompte) took place from June 5 to July 13, 2001 in southern France over the region surrounding Marseille and the Berre Lagoon. The primary objective of this campaign was the study of pollution generation, distribution, and dissipation over this urban and industrial area and the influence of the local atmospheric dynamics and land–sea breeze system forced by a
sharp coastline, rugged orography, and the influence of two large valleys (the Rhône and the Durance). Given the deployment of a very large array of instruments (including surface meteorological stations, wind profilers, lidars, and radiosondes) to monitor both the photochemical and atmospheric dynamical processes during the field experiment, it was decided to associate a GPS experiment with this campaign to share resources and costs. The GPS experiment’s objectives include the determination of the IWV with GPS, the study of the evolution of the IWV in conjunction with the atmospheric dynamics, the study of the GPS-retrieved horizontal gradients, and the development of tomography for three-dimensional restitution of the atmospheric humidity field (Bock et al., 2004). During the campaign, a GPS network of 16 stations has been deployed over the urban and northern border of Marseille within a square area of roughly 20×20 km (Fig. 4). The geometry of the campaign is well suited for tomographic applications due to the good height

Fig. 4. View of the dense GPS network extending over 20×20 km near Marseille.
distribution of the GPS receivers (Fig. 4). To complement the GPS network, a set of additional instruments for the remote sensing of water vapour was deployed. At station VALL, a water vapour radiometer and a solar spectrometer were deployed. At the station called AIRB (Aix-les-Milles airbase), the prototype water vapour Raman lidar of Institut Géographique National was operated. These instruments provide independent IWV measurements for comparison and validation of the GPS solutions.

Processing of the GPS data and preliminary validation of ZTD and gradient parameters are described in Walpersdorf et al., 2004. Using the GAMIT/GLOBK software package (King and Bock, 2000), the processing consists of two successive runs. The first one defines the positions of the GPS stations over 24-h sessions. The second one, tightly constrained positions determined through Kalman filtering, estimates the tropospheric parameters (for each station, one ZTD every 15 min and one gradient every 30 min) over the entire observing period. Thus, following the strategy described in the previous section, we reconstruct the SIWV from the ZTD and gradients provided. However, due to a lack of surface pressure field measurements over the area of interest, we assume no lateral variation of the pressure field on the domain. The hydrostatic gradient could not be estimated reliably and was not accounted for. Nonetheless, the associated error can be estimated by the spatial variation of the ZHD between the only two points of surface pressure measurements available. It is less than 1 mm with a very small temporal variability (less than 0.5 mm) and deemed negligible in this case. Further work will be done with numerical model or with faraway ground pressure measurements to quantify exactly the gradient in the hydrostatic component.

To perform the tomographic inversion using the observed data of the ESCOMPTE experiment, the best results compared to the radio sounding measurements were obtained with a grid size identical to the synthetic case studies for the horizontal (0.05°×0.05°) and with a varying grid size in the vertical from 500 m at the ground (sea level) up to 1000 m at a 10000-m altitude. This offers a compromise between resolution and accuracy, taking into account the increased correlation between the rays at higher altitudes. To validate our results by comparison with radiosonde profiles, we will use the individual 15-min retrievals. In that case, however, 20% of the grid cells are not sampled. It appears that a better strategy to provide denser cell sampling would have been to dispose of 5 min GPS solution interval grouped together during half-hour time to represent a single “time” in the inversion process. It is also to be noted that further increase of the time span would render the inversion unstable because of the high time variability of the water vapour distribution, even in the dry meteorological conditions of our study.

We compare results from day 177 (26 June 2001) at the end of Intensive Observation Period 2b with three soundings launched from the station CINQ in the centre of our GPS network. The launches took place around 12, 14, and 16 h UTC and used Vaisala RS80 sondes. It should be noticed that the GPS profiles extracted from the tomography are mean values of the water vapour over 15-min period and over a large volume of air. Radiosonde measurements are instantaneous and punctual, but the radiosonde ascent takes about half an hour to reach 10,000 m high. Moreover, radiosondes are shifted by the wind. Therefore, when comparing radiosonde and GPS-retrieved absolute humidity profiles, one has to keep in mind these differences. As the two types of measurements are not equivalent, quantitative comparisons are not possible. Fig. 5 shows the comparison between the three
Fig. 5. Water vapour vertical profiles at the GPS station CINQ at 12, 14, and 16 h UTC, 26 June 2001. Red line: initial model. Green line: inverted model. Black line: radiosounding. Numbers on the green line are the diagonal terms of the resolution matrix. The horizontal error bars on the red line are the a priori model error. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)
Fig. 6. Vertical cross-sections at 12, 13, and 14 h UTC, 26 June 2001. Top: north–south cross-section at the longitude of 5.39°. Bottom: east–west cross-section at the latitude of 43.31°. Cross-sections lines are indicated in the small inserted map.
radiosonde profiles (black curves on Fig. 5a, b, and c), the a priori profile used for initialisation (red curves, identical in all panels), and the profile retrieved from tomographic inversion (green curves). The profile retrieved at 12 h UTC is in very good agreement with the radio sounding: it follows well the smooth decrease revealed by the radiosonde, quite different from the a priori profile. However, one could argue that, above 3500 m, the retrieved profile is drawn back to the a priori model which is then close to the actual humidity profile. For the following comparisons, there is less agreement. Indeed, below 3000 m, the agreement between radiosonde and retrieved profile is degraded, while the “spike” in humidity at about 4000 m is missed. A possible explanation, as demonstrated earlier, could be that a thin layer of higher humidity has penetrated above the entire domain between the two soundings. In such a case, the inversion will not be able to resolve the “anomaly” but will “dilute” it over the entire altitude range. That would explain the fact that the “spike” at 4000 m is not reproduced by the inversion and that the retrieved profile below and above presents slightly higher values as the humidity content of the anomaly is distributed at the other levels. This interpretation is supported by the radio sounding measurement at Aix-les-Milles (20 km to the North) which indicates the same spike at 4000 m high.

The final objective of GPS tomographic inversion is the retrieval of the three-dimensional distribution of water vapour over the area of investigation. Such an example is depicted in Fig. 6 where we show successive hourly vertical planes through the studied volume in the N–S and E–W directions from 12 to 14 h UTC. As all the cells of these vertical slices are sampled, there are no possibilities of a water vapour boundary being artificially produced in an unsampled region. In general, the distribution of water vapour is horizontally layered. In the N–S slices, one can observe the appearance of a moist layer moving southward about 1500 m above, what accounts for the urban region of Marseille. Furthermore, a striking feature in the E–W vertical plane is the accumulation of water vapour within the E–W valley from Marseille to Aubagne. The sea breeze would funnel the water vapour along the valley floor, guided by the complex orography. That entire water vapour dynamics is in good agreement with the wind profiler’s data available. The wind profiles show, between 12 and 14 h UTC, a SW sea breeze below about 1000 m and a SE wind above (Delbarre et al., this issue). Although it is still too early in the study to draw definite conclusions on the dynamics involved, this example demonstrates the capabilities of GPS tomographic inversion. One can study the variations of the humidity distribution in conjunction with the local atmospheric dynamics, such as the land–sea breeze effect versus the large-scale circulation.

5. Conclusions and perspectives

Our tomographic software is validated by simulations and comparisons of real data inversions with independent observations. The simulations based on the realistic ESCOMPTE GPS network configuration have shown optimal performance concerning the ray geometry with a ray sampling assuming on 30-min intervals with SIWV values estimated every 5 min. This corresponds to about 200 SIWV observations and only 12% of unsampled grid cells. The optimal grid has a
horizontal step size of $0.05^\circ \times 0.05^\circ$ and 500 m varying vertical cell size. Only a greater number of stations decreasing the mean distance between the stations could allow the use of smaller cells. Places within the model grid where the simulated perturbation was badly resolved can be related to a lack of stations in the realistic network configuration. It should be noticed that, without a priori information, the absolute value of water vapour is less resolved as opposed to relative horizontal variations.

The ESCOMPTE GPS campaign provided an important data set for tomographic inversion. SIWV values, the input data for the tomography, have been obtained by projecting ZWD and horizontal gradient values on the lines of sight of the GPS satellites. No residuals have been used so far for more stability of the tomographic solution. The grid used for the real data inversion differs from the simulation grid by a varying vertical step size of 500 to 1000 m from the surface to a 10,000-km altitude. The inversion is based on a single evaluation of the observables over a 15-min time span. Three inversions have been compared to three successive radiosonde launches in IOP 2b (26 June, 12, 14, and 16 h UTC). Good resolution compared to the a priori model is obtained up to heights of 3000 m. A humidity spike at a 4000-m altitude present in the 14 and 16 h radiosonde data remains unresolved. The reason is probably that the signal is spread homogeneously over the whole network. Such a feature is not resolvable by tomographic techniques.

The temporal aspect of the tomography is not yet implemented in our software. Future work will be done to test Kalman filtering to carry out four dimensional tomography. We will gain some precision in the inversion using the last inversion as the a priori model. The performance could be improved by complementing the GPS measurements with independent water vapour observations (for example, by Raman lidar). Some work is also still needed on the characterization of the GPS residuals. This could be done by a comparison with high-resolution ground-based lidar observations. If the residuals are related to higher order atmospheric structure, they could be included in the inversion and increase the resolution. Furthermore, a finer time resolution could be used so that an anomaly cannot cross the entire domain between two successive inversions. In a global point of view, the continuous increase of GPS permanent stations densifying the existing networks and the addition of the future GALILEO satellite system will enhance the capacities of the GPS water vapour tomography.

In the framework of the ESCOMPTE project, the method will be used to study the influence on the pollution event of land–sea breeze circulation in association with lidar observations (Delbarre et al., this issue) and MesoNH modelling (Bastin et al., 2002). The knowledge of the 3D field of water vapour is therefore essential. The water vapour tomography could also be used in quantitative manner for the validation of high-resolution numerical models.

The capacities of GPS water vapour tomography shown in this study point out some future applications in validating numerical model, inferring water vapour structure, characterizing diurnal cycles and maybe anthropogenic influence, influence of the vegetation, or of soil evaporation/condensation. In the future, the tomographic software will be applied to GPS projects in the framework of Observatoire Hydrométéorologique Méditerranéen Cévennes-Vivarais (OHM-CV), where dense and regional tomography will be compared. In this context, GPS tomography could help to understand the
humidity source for heavy orographic precipitation leading to flash floods observed regularly in southeastern France. The water vapour GPS tomography should be seen as a new tool for meteorologists.

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References


Yuan, L., Anthes, R., Ware, R., Rocken, C., Bronner, W., Bevis, M., Businger, S., 1993. Sensing climate change using the global positioning system. J. Geophys. Res. 98 (D8), 14925–14937.