

# Incorporation of InSAR and LandSat ETM+ with CGPS and SVM to assess the subsidence of Mexico City

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## Abstract

This study presents an enhanced analysis of the subsidence rates and their effects on Mexico City. As a result of excess water withdrawal, Mexico City is experiencing subsidence. To analyze this subsidence of Mexico City, we integrated and analyzed Interferometric Synthetic Aperture Radar (InSAR), Continuous Global Positioning Systems (CGPS), and optical remote sensing data. This study utilized 52 ENVISAT-ASAR, nine GPS stations, and one Landsat ETM+ image from the Mexico City area to gather a better understanding of the subsidence rates and its effects on Mexico City's community. The InSAR data covers the period between March 2002 until June 2010, and the GPS data span the period from 1998 until 2012. We find that the maximum of 352-mm change in Line Of Sight (LOS) direction supports the outcomes from previous studies. This study shows that the maximum rate of subsidence in Mexico City is 352 mm/yr. The finding of this study reveals a high amount of correlation (up to 0.98) between two independent geodetic methods. We also implemented the Support Vector Machine (SVM)

analysis method based on Landsat ETM+ image to classify Mexico City's population density. We used SVM to compare Persistent Scatterer Interferometry (PSI) subsidence rates with the buildings' distribution densities. This study improves the existing method by incorporating 52 ENVISAT images, ETM+, SVM classification, and 9 CGPs. This integrated study shows that the fastest subsidence zone (i.e. areas greater than 100 mm/yr), which falls into the above-mentioned temporal baseline, occurs in the areas of high and moderate building distribution density.

**Keywords:** Mexico City; Subsidence; InSAR; GPS; PSI; SVM

## 1. Introduction

Interferometric Synthetic Aperture Radar (InSAR) data have been available for geological and geomorphological analysis since the launching of ERS 1 in 1992. Persistent Scatterer Interferometry (PSI) has been employed for more than 15 years to monitor the surface of the Earth. In the last two decades, PSI and similar techniques have been proposed as well ([Usai and Hanssen 1997](#); [Hanssen 2001](#); [Ferretti et al. 2001](#); [Lanari et al. 2004](#); [Salvi et al. 2004](#); [Ferretti, et al. 2011](#); [Poreh et al. 2017](#)). Developed by researchers at the Politecnico di Milano (POLIMI), the procedure of the PSI technique is known as Permanent/Persistent Scatterers Interferometry ([Ferretti et al. 1999](#); [Wang et al. 2010](#)). Prior to monitoring similar terrain subsidence, several researchers used geodetic methods such as InSAR, PSI, and Continuous Global Positioning System (CGPS) ([Berardino et al. 2002](#); [Mora et al. 2003](#); [Crosetto et al. 2005](#); [Hooper 2008](#); [Crosetto et al. 2008](#); [Pepe et al. 2011](#); [Perissin and Wang 2012](#); [Navarro-Sanchez and Lopez-Sanchez 2013](#)). A large number of radar images and techniques were considered for the estimation of historical changes on the Earth's surface. For example, similar methods, such as the Small Baseline Subset (SBAS), have emerged with the effectiveness of the PSI approach ([Murillo and Manuel 1995](#); [Rudolph et al. 2006](#); [Gourmelen et al. 2007](#); [Dai et al. 2016](#)).

The subsidence in Mexico City (Fig. 1), which is due to the extraction of groundwater, began in the 1840s. Later, this phenomenon became extensive in the 1930s and 1950s (Carrillo 1947; Ortega-Guerrero et al. 1999; López-Quiroz et al. 2009; Rodell et al. 2009; Lopez-Quiroz et al. 2009; Ayazi et al. 2010; Ortiz-Zamora and Ortega-Guerrero 2010; Osmanoglu et al. 2011; Yan et al. 2012; Chaussard et al. 2014; Castellazzi et al. 2016; Albano et al. 2016; Castellazzi et al. 2016; Suárez et al. 2018; Cigna et al. 2019). Mexico City is built on highly compressible clays (Cuevas 2004). The mechanism of the subsidence in Mexico City lacks enough natural water recharge (i.e. no extracted water replacement); consequently, further compaction of the clay layers is expected (Carrillo 1947; Cuevas 2004; Osmanoglu et al. 2011; Albano et al. 2016). The subsidence is not associated with the aquifer. Instead, this subsidence may exist because of the compaction of the overlying geological layers (Castellazzi et al. 2016; Suárez et al. 2018; Cigna et al. 2019).

Fig. 1 a) Shaded relief map of Mexico City overlaid by topography map and installed CGPS stations in the metropolitan area. The white rectangle shows the ENVISAT-ASAR data coverage area in the descending mode (see Fig. 2). b) 3D view of Mexico City area with Landsat ETM+ imagery from 25/11/2005. c) Cross-sections A-A' from SRTM topography data (Source: <http://www.glcf.umd.edu> (Cabral-Cano et al. 2008))

In Mexico City, subsidence occurs when irrecoverable aquifer systems are heavily pumped. This operation results in the loss of aquifer storage and damage to engineered structures. Since the late 1950s, this subsidence has accelerated to a remarkable extent, and considerable correlated structural damage has been reported in the Mexico City area (Cabral-Cano et al. 2008; Osmanoglu et al. 2011; Yan et al. 2012). More than 20 million inhabitants in the metropolitan area are facing this extraordinary land subsidence hazard (<http://www.unesco.org>). For example,

in Mexico City Metropolitan Cathedral (which took 250 years to build), one side is settled nearly 2.44 meters deeper than the other side. Furthermore, the cathedral is leaning to the left side (<http://www.unesco.org>).

[Cabral-Cano et al. \(2008\)](#) studied Mexico City's subsidence based on the InSAR methodology. By using ERS and ENVISAT-ASAR satellite data in the temporal baseline of 1996-2003, the authors pointed out that the subsidence rate in Mexico City has reached 370 mm/yr. It is controlled by compaction of the quaternary lacustrine clays and silts. [Castellazzi et al. \(2016\)](#) studied the land subsidence in major cities in the whole country including Mexico City. [Suárez et al. \(2018\)](#) also estimated the ground subsidence in the city of Morelia, Mexico by applying InSAR. [Cigna et al. \(2019\)](#) surveyed the wide area of surface deformation in urban areas and geothermal fields in the eastern Trans-Mexican Volcanic Belt, Mexico by utilizing InSAR techniques.

Fig. 2 Geological map of the study area showing Mexican Volcanic Belt (MVB) and other formations (black frame refers to [Fig. 1](#) ground coverage) (Source:

<https://pubs.usgs.gov/of/1997/ofr-97-470/OF97-470L/graphic/data.htm>

[Castellazzi et al. \(2016\)](#) used GRACE and InSAR data sets to remotely assess groundwater storage loss in the Mexico City area. Using the SBAS-InSAR algorithm to reveal areas subject to ground motion related to groundwater overexploitation, they noted that GRACE satellite data sets do not entirely detect the significant groundwater losses. As a solution, these data sets should be combined with other high-resolution satellite imageries. [Chaussard et al. \(2014\)](#) used ALOS data to retrieve the deformation rate in Mexico City for 2007-2011. They identified land subsidence in 21 areas, including 17 cities, and observed that the maximum rates in Mexico City are 300 mm/yr. [Ortiz-Zamora et al. \(2010\)](#) used a ground magnetic survey combined with lithologic logs to map the extension of basalts, ground surface data, and hydraulic data. The

authors ran predictive simulations and field data to predict the deformation rates in the Mexico City area from 1984-1989. [Ortega et al. \(1993\)](#) used hydraulic data from a network of monitoring wells, geotechnical data from core samples, and historical information for 1984-1989. The authors ran a mathematical model to predict future subsidence under the current pumping rates.

[Strozzi and Wegmüller \(1999\)](#) used ERS satellite data sets in the temporal baseline of 1995-1997 to monitor subsidence in Mexico City. They found 400 mm/yr subsidence in the eastern part of Mexico City's community. They concluded that because of water extractions, almost nine meters of subsidence has occurred in the Mexico City area over the last century. [Yan et al. \(2012\)](#) compared PSI (with Gamma-IPTA chain methodology) and SBAS for Mexico City's subsidence from 2002-2007 and explained the strengths and weaknesses of each method. [Lopez-Quiroz et al. \(2009\)](#) analyzed 38 ENVISAT images acquired between 2002 and 2007 and discovered 400 mm/yr subsidence with InSAR. They also showed that the deformation is almost linear over time baselines. A PSI study based on ENVISAT-ASAR data was carried out in the eastern part of Mexico City during a short temporal baseline (i.e. 2004-2006) by [Osmanoglu et al. \(2011\)](#), [Castellazzi et al. \(2016\)](#), and [Cigna et al. \(2019\)](#). They used only 23 images for PSI data analysis, which is the minimum number of sufficient imageries for PSI analysis. Finally, [Sowter et al. \(2016\)](#) used Sentinel for temporal baselines of 2014-2015. The SBAS (ISBAS) technique was utilized to measure the deformation rate of 240 mm/yr along the LOS direction, equivalent to over 400 mm/yr vertical rates ([Zebker et al. 1997](#); [Cigna et al. 2019](#)).

However, despite the controlling procedure and the above-mentioned research, we used ten years of InSAR data in this study to focus on the maximum 352 mm/yr displacement rate (in LOS direction) occurring in the central and eastern parts of Mexico City. Our study covers a longer temporal baseline (between November 2002 and June 2010) and a wider area ( $62 \times 56 \text{ km}^2$ ) than

the previous studies. In this study, we analyzed 52 ENVISAT-ASAR data from November 2002 until June 2010 with a coverage area that is twice the size as that used in Osmanoglu's work (Osmanoglu et al. 2011). We managed to extend the temporal baseline. As pointed out by Zebker et al. (1997), for  $N$  independent interferograms ( $N+1$  InSAR imageries), the temporally uncorrelated noise reduces by a factor of  $1/\sqrt{N}$ . Therefore, this statement supports the current study while enhancing the reliability performance of similar works (Cabral-Cano et al. 2008; Osmanoglu et al. 2011; Albano et al. 2016).

In summary, the objectives and advantages of this study as compared to the existing research are (1) analyzing more InSAR imageries (longer temporal baselines), (2) working with a larger coverage area, (3) using accurate ENVISAT images in conjunction with more CGPS stations, and (4) conducting a detailed geohazard risk assessment of Mexico City based on SVM integration.

## 2. Geological and geotechnical settings

When Spanish invaders conquered North America in 1521, they built Mexico City over the ruins of the Aztec civilization capital of Tenochtitlan. The old Aztec city was an island in Lake Texcoco (Fig. 1 and Fig. 2). The Spanish drained the lake over an extended time and expanded Mexico City onto the new land, where it exists today. Almost the entire city stands on layers of sand and clay with thicknesses up to 300 meters at some locations.

These soft, water-laden and loose sediments make the city uniquely vulnerable to subsidence, earthquakes, and other kinds of geohazards. The 100-km long and 80-km wide NE-SW-oriented Mexico Basin is located in the eastern sector of the Mexican Volcanic Belt (Fig. 2). This volcanic zone is the result of the subduction of the Cocos and Rivera oceanic plates underneath the North America plate. Morphologically, the Mexico Basin includes (a) volcanic ranges,

composed of either polygenetic or monogenetic volcanoes; and (b) a series of fan-like knolls, located at the base of each volcanic range. The intercalation of pyroclastic and epiclastic deposits and (c) flat-land areas results from the accumulation of lacustrine sediments of variable thicknesses interbedded with tephra layers (Ferrari et al. 1999; Arce et al. 2013).

The lacustrine sediments of Mexico City exhibit very unusual behaviour; in some cases, the water content can surpass 500%, the plasticity index sometimes exceeds  $I > 300\%$ , and the compression index could be in the order of 10 (Castellazzi et al. 2016). The friction index of Mexico City clayey soils is comparable in magnitude to that of sands. Furthermore, the sediments do not show strength loss, not even when the amplitude of the cyclic loading is as high as 80%. The sediments of Mexico City are a complex mixture of crystalline minerals and amorphous material with heterogeneous volcanic and lacustrine sediments (silty clay or clayey silt) (Ortega-Guerrero et al. 1993, Du et al. 2019).

### 3. Data and methodology

#### 3.1. Data gathering

Fig. 1 shows the study area in Mexico City with CGPS station locations and InSAR data coverage area. We used complementary data sets from nine GPS stations (daily data), thousands of observations of Permanent Scatterers (PS) from ENVISAT-ASAR satellite, and optical remotely-sensed imageries. The deformation signature of the Mexico City area with temporal baselines of 1998-2012 and 2002-2010 from GPS and InSAR data, respectively, provides updated estimates including related risk assessments. For some GPS stations, we do not have data from 1998, and some of the data have gaps and are of poor quality (see Fig. 3). These nine GPS stations were installed and maintained by the University of Mexico. We used CGPS as

ground truth and calibration tool for the InSAR data. With more than six years (and eight years for one station) of CGPS and InSAR overlap, the two independent geodetical methodologies provide an updated picture of subsidence in Mexico City and its surrounding area in the first decade of the 21st century.

This study used the Landsat ETM+ image to classify the types of man-made structures and buildings' densities in the study area. Landsat ETM+ image with seven bands, acquired on 15/11/2005, was processed by the U.S. Geological Survey (USGS). We used DORIS for SVM analysis incorporation with InSAR/PSI processing techniques. We also applied supervised classification using ENVI 4.8 under Windows and Linux platforms for image processing.

### 3.2. GPS

All of the CGPS stations are inside the ENVISAT-ASAR data coverage (white frame in [Fig.1](#)). UCHI, UGOL, UIGF, and UGAL are located either on the andesite-basalt lava or on the tephra deposits around Mexico City ([Osmanoglu et al. 2011](#)). CGPS data was provided by the University of Mexico and has been analyzed by utilizing the precise point positioning of the ITRF-2000 reference frame ([Dixon et al. 2000](#); [Altamimi et al. 2002](#); [Hilley et al. 2004](#); [Argus 2007](#)). The GPS data have been converted from the ITRF coordinate system to the north, east, and vertical coordinates utilizing the national coordinate system. The comparison of PS with the GPS data is attempted, and the velocities are compared to a reference point. In the Mexico City area, the tectonics (i.e. plate movements) also plays a very important role, which has an undeniable effect on the InSAR reflectors, including PS's reference point. In other words, the selection of reliable coordinate systems is crucial for the fusion of InSAR and GPS.

Fig. 3 North, east, and vertical components of motion observed by the permanent GPS stations (blue points) using point positioning and the ITRF-2000 reference frame. GPS rates are provided in Table 1

This study calculated velocity and uncertainty for each CGPS station by linear regression analysis. For details of the procedures and estimation of uncertainties for GPS measurements, refer to [Dixon et al. \(2000\)](#) and [Sella et al. \(2002\)](#). The results of the CGPS data analysis are provided in [Fig. 3](#) and [Table 1](#). These data are presented in “absolute” coordinates (latitude, longitude, and height). All of the vertical (UP) components show negative values (subsidence) with a maximum of  $-275.3 \pm 3.5$  mm/yr. It identifies subsidence in the MRRA station or the easternmost CGPS station ([Fig. 1](#)). A minimum of  $-0.3 \pm 2.5$  mm/yr is observed in the UCHI station, which is located on the andesite-basalt lava or tephra deposits (see [Fig. 1a](#) and [Fig. 2](#)). GPS data could be contaminated with several potential noises. Because the study area is small, orbital effects are assumed to be spatially uniform. Despite different temporal coverages of CGPS and InSAR data, we have managed to compare these two independent geodetic methods. For comparison of GPS and PSI data, GPS-observed data should be aligned with to the PSI points. The velocities  $V_{\text{LOS}}$  measured by the PS technique are in the Line-Of-Sight (LOS) direction, and the GPS measurements are projected to this direction via:

$$V_{\text{LOS}} = V_{\text{East}}\psi_{\text{East}} + V_{\text{north}}\psi_{\text{north}} + V_{\text{Vertical}}\psi_{\text{Vertical}} \quad (1)$$

Where each  $\psi$  component is known as a directional cosine  $\psi = [\psi_{\text{East}}, \psi_{\text{north}}, \psi_{\text{Vertical}}]$ . For example, LOS’s vertical component has a directional cosine of  $\cos(\theta^\circ)$ . Therefore, with only vertical movement, the LOS velocity should be divided by  $\cos(\theta^\circ)$  to

represent the exact amount of the vertical component of motion. It should be considered in the LOS interpretations. Regarding this setting, we can infer LOS data as a vertical displacement with an error rate of almost 8% ( $\cos(23^\circ) = 0.92$ ) for ENVISAT-ASAR satellites. Despite a relatively small amount of horizontal deformation rates in the Mexico City area (based on CGPS data) in comparison with a vertical deformation rate, we consider PSI velocities as a combination of vertical and horizontal velocities (based on equation 1).

Table 1 Deformation rates of CGPS stations in the Mexico City area (point positioning solution, ITRF-2000 networks). For the location of CGPS stations, see [Fig. 1](#). For geology, see [Fig. 2](#).

### 3.3. Interferometric SAR and PSI

In this study, we used 52 ENVISAT-ASAR scenes to analyze the subsidence in the study area. The study area covers a temporal baseline between November 2002 and June 2010. The 05/05/2006 acquisition has been selected as the master scene to minimize the effects of spatial and temporal baselines ([Zebker et al. 1997](#); [Scharroo and Visser 1998](#); [Hanssen and Bamler 1999](#)). The ENVISAT-ASAR imageries, which cover an area of  $62 \text{ km} \times 56 \text{ km}$ , are centered over Mexico City's historic downtown. In the first step, we ran the crop by applying Delft Object-oriented Radar Interferometry Software (DORIS) method. Next, we oversampled the data by using a factor of two in range and azimuth to avoid any undersampling of the interferograms, especially during the resampling of the slave acquisition ([Bell et al. 2002](#)). By running this step, we prevented aliasing over the data. In the interferometry step, scenes were oversampled by a factor of two in range and azimuth to make each pixel (originally  $4 \text{ m} \times 20 \text{ m}$ ) approximately  $2 \text{ m} \times 10 \text{ m}$ . DORIS was used to make differential interferograms of imageries ([Hanssen and Bamler 1999](#)). For reducing the orbital effect in the produced interferograms, we used precise orbit data (DOR and VOR) from the TU-Delft University that were supposed to minimize orbital errors

significantly ([Scharroo and Visser 1998](#)). We appended useful orbit data to the images. We did not run the Porbits step (in DORIS software) in our data.

**Table 2** illustrates the baseline information (perpendicular, temporal, and Doppler baselines) for 52 ENVISAT-ASAR satellite imageries. We acquired images in the descending mode for the study area. Imageries are given in the format of YYYYMMDD (first and fifth columns). Major factors influencing the InSAR Phase measurements are ([Hanssen 2001](#)):

$$\emptyset = \varphi_{\text{atm}} + \varphi_{\text{orb}} + \varphi_{\text{def}} + \varphi_{\text{scat}} + \varphi_{\text{DEM}} + \varphi_{\text{n}} \quad (2)$$

$\varphi_{\text{def}}$  is the part connected to deformation,  $\varphi_{\text{DEM}}$  is the topographic phase contribution, and  $\varphi_{\text{orb}}$  is the orbital part error that could be minimized by using precise orbital data.  $\varphi_{\text{atm}}$  is the atmospheric phase screen, and  $\varphi_{\text{scat}}$  is the change in the scattering attributes of the scatterers during the time that may not fall in the urban area. Finally,  $\varphi_{\text{n}}$  is the noise part of the phase, which is for strong scatterers and would be negligible. Data analysis and probable errors are explained in [Hanssen and Bamler, \(1999\)](#), [Hanssen \(2001\)](#), and [Kampes \(2005\)](#).

The georeferencing accuracy of standard ENVISAT-ASAR images concerning reference ellipsoid is approximately 12 m in azimuth and 60 m in range direction. To obtain as high as possible level of accuracy in geolocation, SRTM 3 arcsec data have been used to the scale of 10 and 15 m in azimuth and range direction, respectively. **Fig. 4** shows the ENVISAT-ASAR data configuration used in this study. This figure depicts the temporal and perpendicular baselines of the ENVISAT-ASAR data for the study area in Mexico City. Scene 20060505 has been selected as a master image to reduce the orbital error and maximize the coherence. In **Fig. 5**, interferograms are presented. Interferograms are not geo-referenced (in radar coordinates). Each color cycle (or fringe) shows a movement in the rate of 2.83 cm in LOS direction.

We applied the TU-Delft approach for analyzing the data and determining the average subsidence rates, including the deformation time series for each pixel on the ground (Ferretti et al. 1999; Wang et al. 2010; Albano et al. 2016; Poreh et al. 2017; Du et al. 2019; Cigna et al. 2019). CGPS stations in the area cover SAR's data and the nearby region. It was employed to compare InSAR data results with the CGPS data for calibrations and ground truth controls. This study used the phase unwrapping method to solve the phase history of a single-pixel called periodogram (Osmanoglu et al. 2011). As pointed out in previous studies (Cabral-Cano et al. 2008; Osmanoglu et al. 2011), the subsidence in Mexico City is almost linear in time. Therefore, the most accurate model appears to be the linear model. Note that in PSI data analysis philosophy, non-linear displacements between neighboring scatterers are negligible. Nonetheless, we applied different deformation models and statistical hypothesis tests in unwrapping the steps to minimize the phase ambiguity errors. The reference height for the PSI analysis is fixed to 2240 meters to reduce the topographic phase errors efficiently. We used the PSC selections method and considered the threshold of dispersion index with the selection of a 300-meter fixed grid size.

Table 2 Baseline information of ENVISAT-ASAR imageries

Note: B.Temp is the temporal baseline (time difference) between master and slave acquisitions, B.Perp is the perpendicular baseline between orbits of the master and slave scenes, and B.Dopp is the difference between the Doppler centroid frequencies of master and slave scenes.

Fig. 4 ENVISAT-ASAR data sets configuration

Fig. 5 Differential interferograms (from 51 interferograms) for Mexico City study area. Images

(a)-(i) are the interferograms between the master image and radar images acquired on 2003/04/11, 2003/08/29, 2004/09/17, 2004/11/26, 2004/12/31, 2005/07/29, 2005/09/02, 2005/11/11, and 2007/03/16, respectively. (For characteristics of each image see [Table 2](#))

### 3.4. Building density classification

In this study, we used the Landsat ETM+ image (with seven bands) dated 15/11/2005. We collected the satellite image from the University of Maryland web page [41]. We classified the extension of populated buildings from the image. We used the False Color Composite (FCC) (2:3:5) to generate the RGB image. This study applied the SVM method on the RGB image to classify the populated buildings from the study area and to give a sense of the densities of the buildings ([Fig. 6](#)).

SVM is a well-known classification method that currently has wide applications in image processing and machine learning. This SVM statistical method is designed for recognizing different patterns, classification, and regression analysis. In the simplest form, SVM classification finds a hyper plan such as a  $\forall i \ [y] \_i (x\_i.w+b)-1$  problem that could segregate two classes by the minimization of the following Lagrange equation ([Hilley et al. 2004](#)):

$$\frac{1}{2} \|\mathbf{w}\|^2 - \sum_{i=1}^L \alpha_i y_i (\mathbf{x}_i \cdot \mathbf{w} + \mathbf{b}) + \sum_{i=1}^L \alpha_i \quad (3)$$

Where  $\forall i, \alpha_i \geq 0$ , and  $L$  are the number of training points, and  $\mathbf{w} = \sum_{i=1}^L \alpha_i y_i \mathbf{x}_i$

Fig. 6 SVM classification of the Mexico City area based on the Landsat ETM+ image from 15/11/2005 (red frame in Figure 1 refers to SVM ground coverage)

In this study, we used three training sets to classify the patterns. We considered three different kinds of building density classes as the training sets. We selected three areas with three Regions Of Interest (ROI) to apply supervised SVM classification analysis. These three areas include (i) highly populated areas (dense building distribution), (ii) sparse building distribution with less density, and (iii) areas that are mountainous and unpopulated. After selecting these three ROI, we imposed the SVM method to classify these areas mentioned above with minimum error.

#### 4. Results and discussion

The ability to make combined measurements from InSAR, GPS, and optical remote sensing imageries may be a powerful tool for studying the subsidence in Mexico City and the related risk management. The combination of CGPS and PSI methods seems straightforward. However, this process requires careful analysis since we do not have radar reflectors (PSs) in the exact location of each CGPS station. Therefore, the PSI approach is a relative method in comparison with the absolute GPS methodology.

#### 4.1. CGPS

As pointed out by [Osmanoglu et al. \(2011\)](#) and [Cabral-Cano et al. \(2008\)](#), the eastern and central parts of Mexico City (i.e. the most populated areas in the Mexico City community) have undergone a high degree of subsidence. [Table 1](#) depicts the CGPS results of the displacement in Mexico City, and [Fig. 3](#) illustrates the extraordinary subsiding in the central and eastern parts of the Mexico City community. All CGPS stations show negative values (subsidence) in UP (vertical) components ([Fig. 3](#)). This subsidence contains the regional tectonic movements and local subsiding because of the water withdrawals. The natural compaction of basins is causing slow subsidence in rates of a few millimeters per year in areas like Mexico City ([Ortega-Guerrero et al. 1993](#)). In comparison, subsidence based on groundwater extractions has rapid rates of ten centimetres per year ([Nelson 2000](#); [Cabral-Cano et al. 2008](#)). For almost all of the CGPS stations in the study area, seasonal variations in UP components are negligible. The GPS data shows high agreements with the previous studies in this area ([Cabral-Cano et al. 2008](#); [Osmanoglu et al. 2011](#)).

The permanent GPS station UIGF (see [Fig. 3](#)) was installed in 1998 in the Mexico City metropolitan area. MOCS, MPAA, and MRRA stations have been continuously recording the subsidence since 2005 with high temporal resolutions. These CGPS stations, which are located in the high subsidence region, record the vertical subsidence in the range of  $-164.7 \pm 2.8$  and  $-275.3 \pm 3.5$  mm/yr. The highest vertical subsidence in the study area belongs to the station MRRA with a rate of  $-275.3 \pm 3.5$  mm/yr. We recorded  $-2.2 \pm 2.7$  mm/yr vertical subsidence for UIGF station. The station UPEC, located farther to the west (see [Fig. 1](#)), shows vertical subsidence in rates of  $-82.3 \pm 0.7$  mm/yr with a negligible amount of seasonal variations. Some of the other

CGPS stations show poor and unreliable data distributions. For instance, the stations UTEO, UJAL, and UCHI show lots of gaps and missing data.

#### 4.2. InSAR and PSI

In the InSAR and PSI techniques, the displacements are resolved in the Line-Of-Sight (LOS) direction. This study is not in the three orthogonal displacement vectors; therefore, the combination with CGPS stations must be dealt with very carefully. Fig. 7 depicts the PSI rates in the Line-Of-Sight (LOS) direction for the subsidence in Mexico City during 2002-2010. Most of the available PSs time series do not address the issue of seasonal variability; thus, the amount of aquifer recharge is weaker. The rate of the PSI is increasing eastward and showing a massive amount of movement towards the center of the basin where the clay-rich sediment package is thickest (see profile A-A' in Fig. 1a).

Fig. 7 shows subsidence through the eastern and central parts of the city (from November 2002 until June 2010). The maximum amount of 352 mm/year displacement in LOS direction occurs in the eastern part of the city toward the remnant of Lake Texcoco (see Fig. 1). PSI rates are shown in Figure 7, and nine GPS locations overlaid on the shaded relief map of the STRM 3 arcsec data (<http://www.glcf.umd.edu>) (Fig. 7). For each pixel, a time series of displacements in the temporal baseline (November 2002-June 2010) has been plotted. These are used later to compare GPS with PSI data for calibration and validation.

Fig. 7 PSI deformation rates of the Mexico City area for the temporal baseline of 2002-2010

Table 3 Comparisons of LOS' rates of movements for GPS and PSI

**Table 3** illustrates the deformation rates of nine GPS stations inside the ENVISAT-ASAR data coverage area (see Figure 1a) and the closest PSI rates to those stations. It shows that the PSI and the CGPS rates are in agreement (see Fig. 9). Moreover, to summarize, more than 600,000 points have been selected as permanent scatterers in the study area. For these points, the authors calculated time series (2002-2010) and the height of radar scatterers. The PSs density in the study area is 240 PSs/km<sup>2</sup>. Fig. 8 shows the histogram of LOS deformation rates, which is strongly skewed to negative values. The J-shape distribution of the deformation rates is linked to the record of a strong subsidence pattern in Mexico City. The cause of deformation is probably due to groundwater extraction and compaction of the overlying geological layers. Other factors, such as tectonic movements of the plates which are mostly in the north direction, have a small effect on the existing PSI rates. The most common deformation rates are between -20 and +2 mm/yr. The subsidence is more visible when the geological composition of the underground is associated with the fluvial clay sequences and, of course, not to the volcanic rock series. A comparison with previous InSAR and PSI works shows a high amount of agreement with the shape and the rate of the subsidence (Cabral-Cano et al. 2008; Osmanoglu et al. 2011).

Fig. 8 Histogram of PSI's deformation rates for Mexico City study area

#### 4.2.1. Comparison of PSI with CGPS stations and validation

The PSI deformation rates measure the time interval between the generated interferograms, while the CGPS stations measure in the constant rate of daily data. However, more than six (and eight for one station) years of overlaps can help us to compare CGPS stations with PSI data. Despite the poor quality and gaps for some CGPS stations (for instance UTEO and UJAL), data from other CGPS stations could be compared with the PSI-derived deformation rates.

The accuracy of the PSI method can be measured in sub-centimetres if a sufficient number of imageries are utilized in PS analysis (Ferretti et al. 2011). We compared the outcomes of this research with nine independent CGPS stations to assess the accuracy of the PSI. We found that there is an improvement of the previous studies. Table 2 lists the LOS rates obtained from point positioning GPS and PSI analysis. For each GPS station, a search was conducted to find the closest PSI point's rates. PSI points close to the CGPS stations show similar rates of subsidence in the Mexico City metropolitan area. Fig. 9 shows the graph interpretation of Table 2. A one-to-one line shows perfect agreement between GPS and PSI points, leading to acceptable PSI analysis in the Mexico City study area.

Because of the low incidence angle ( $\sim 23^\circ$ ) of the ENVISAT-ASAR imageries, most of the PSI recorded displacements are from vertical movements of the terrain. This measurement justifies the perfect agreement between two independent displacement monitoring tools (PSI and CGPS). Furthermore, recorded deformation rates from independent CGPS stations of MOCS, MPAA, MRRA, and UPEC confirm that the horizontal movements are small and do not follow a preferred direction. For instance, MPAA and UPEC are horizontally moving in the northwest direction. MOCS is moving horizontally almost to the north, and MRRA is moving in a southwest direction. On the other hand, a comparison with PSI histogram (Fig. 8) shows that extraordinarily high rates of displacements (rates less than  $-20$  mm/yr) are occurring within the Mexico City metropolitan area. Since the two independent methods, PSI and CGPS, with different time intervals are in agreement, the seasonal behaviors are negligible.

Fig. 9 GPS and PSI comparison in the Mexico City study area from Table 2. The dashed line is 1-to-1 line shows great correlation of the GPS and PSI data.  $R^2$  is in the range of 0.98

#### 4.2.2. LOS GPS and average PSI comparison

For each CGPS station, we selected a subset of PSs located nearby to the CGPS stations to test the stability of CGPS stations with their surroundings. The comparison of CGPS and PSI displacement results is illustrated in [Fig 10](#). For this purpose, we selected CGPS stations MOCS, MPAA, and MRRA, which were installed after 2004 and are located in the zones where the subsidence is more visible. The MPAA and MRRA show an obvious acceleration in the north and/or east directions (see [Fig. 3](#)), and the station MOCS also has a small amount of acceleration in the horizontal directions. However, the main displacements are in the vertical directions. In order to tie the PSI results to CGPS data, this study projected the north, east, and vertical directions to the LOS direction (via equation 1) with directional cosine. Next, the average of PSs displacement rates in distance of  $r = 1, 2, 4, 6,$  and  $8$  km were examined. In all of the tested CGPS stations, for  $r > 2$  km, the deviation is remarkable and the stability would be poor. Therefore, we selected the maximum distance from each CGPS station in  $r = 2$  km. Statistics of the selected PSI points around the GPS antennas MOCS, MRAA, and MPAA are given in [Table 4](#). This table depicts the number of selected PSs, Min, Max, Mean, and SD for each GPS station.

Table 4 Statistics of the selected PSI points around the GPS antennas

Fig. 10 Comparison of LOS GPS and average of subsets from PSI method

For three CGPS stations, CGPS data was projected in the LOS direction by Equation 1. Averaged PSI rate in subsets for  $r = 1, 2$  km was calculated. As we had seen in the previous sections, for MOCS and MRRA, average PSs rates after 2 kilometers deviate from CGPS displacement rates. MPPA stays more or less stable to the distance changes.

For station MPAA, which is the southernmost CGPS station of the three stations above, a perfect correlation has been observed. Until there is a maximum 2-km distance from MPAA, this CGPS station has valid results. A small amount of deviation from this CGPS station after 2009 (June) has been observed. For CGPS stations MOCS and MRAA, the correlations are good until 2007. After 2007, for both stations in  $r = 1$ , 2-km small deviations have been observed.

The immediate observations from this analysis are as follows: (a) the slopes for both CGPS and average PSI are similar and lead to the same amount of subsidence rate with two independent methods; and (b) The effective and valid area for each CGPS station is approximately 2 km. In other words, we can trust the GPS and PSI correlation for a circle with a radius of maximum 2 km.

### 4.3. Risk assessments for existing subsidence in Mexico City

The comparison of Mexico City's subsidence monitoring history with the existing research and current study is depicted in [Table 5](#). This study used larger InSAR and more CGPS temporal baselines in conjunction with the combination of other available remotely-sensed data. We used the SVM classification based on Landsat ETM+ imagery. We applied SVM to analyze the populated area in Mexico City and the comparison with the subsidence rates from PSI data as well as risk assessment. The RGB composite (2:3:5) of three available bands was generated, and then we oversampled the composite image with three ROIs. In this last stage, the SVM classifier was applied to the image to obtain the buildings' densities. [Fig. 6](#) illustrates three classes of densely populated area in Mexico City. We overlaid the CGPS stations on the SVM classification map to compare this map with the terrain displacements maps. Comparison with the PSI deformation rates stress that the denser building zones are located on the highest

deformation zones. This comparison also stresses that areas with a high population are at risk for a high degree of subsidence in the eastern and central parts of Mexico City.

In the Mexico City area, the ground deformations are classified into (1) sudden subsidence and (2) slow subsidence. In general, slow subsidence usually causes enormous economic and human-related disasters. Cities built on unconsolidated clays, silts, peats, or sands are in danger of sudden or slow subsidence. Extreme groundwater extractions, flooding, tsunamis, and an earthquake have made the morphological settings in Mexico Basin be a dangerous situation of subsidence. In particular, buildings and streets add weight to the region and intensify the soil's stress even more. Meanwhile, finding the buildings' extensions and their relationship with ongoing subsidence is crucial.

#### Table 5 Mexico City subsidence monitoring history

The existing subsidence due to over-pumping in the Mexico City metropolitan area has been examined in this study. [Nelson \(2000\)](#) showed that a maximum of nine meters of subsidence in an area as large as 225 km<sup>2</sup> had been observed in the metropolitan area. As mentioned previously, the main subsidence occurs because of water extraction and the consequent compaction of the alluvial sediments. In the subsiding area, the intergranular pressure of aquifers decreases. The depletion of water at depth is the cause of the observed subsidence. The highest subsidence rate is located in the central and eastern parts of Mexico City. The area experiences rapid subsidence in regions with a subsidence rate greater than 50 mm/yr ([Fig. 7](#)). It correlates with the regions of intense groundwater extraction such as Texcoco sediments ([Fig. 1](#)) ([Osmanoglu et al. 2011](#)). The more stable area is located on the western side of Mexico City (in the mountains). As is evident from a comparison of [Fig. 6](#) and [Fig. 7](#), the highest deformation

rates are located in the region with high densities of buildings. As the subsidence is developing continually in this region (see GPS results in [Fig.3](#)), the city's central and eastern parts are threatened. With a subsidence rate of 352 mm/yr, there will be a total of 3.5 meters of displacement in ten years. In this case, floods in rainy seasons will be of great concern in these two parts of the city. However, this may be an inaccurate estimation. Because the compaction of aquifers in response to water extraction depends on the aquifers' physical properties subsidence may slow down over time ([Terzaghi 1925](#)). With more than 2500 m of sediments in Mexico Basin, estimating the stopping point of subsidence is difficult ([Jose, R. and Sanchez 1989](#); [Scharroo and Visser 1998](#)). We need additional subsurface data for more accurate estimations analysis. The weak subsoil, which is mostly made up of elasto-plastic clays minerals has the capacity for instantaneous and high compressions. Meanwhile, the results of exceeding water withdrawal have made the Mexico City area susceptible to any geohazard threats. For instance, in 1985, at least 40,000 people died in an M 8.1 earthquake ( $18.2^{\circ}$  N,  $102.7^{\circ}$  W) with an epicentre 350 km away from Mexico City. The damage was directly related to amplification phenomena due to local stratigraphic settings, and the central part of Mexico City suffered an average vertical displacement of 30 cm ([Murillo and Manuel 1995](#)). Economically, the costs associated with the subsidence are enormous and include more than three billion USD due to the immediate collapse of 412 buildings and the severe damage of another 3,124 buildings ([Murillo and Manuel 1995](#)). Maintenance and geotechnical supports have indirectly led to the increase in flood risk, soil fractures, and other threats to human life. As these costs grow over time, it becomes increasingly important to assess the extension and magnitude of potential damage. Meanwhile, the monitoring of subsidence must be continued with more GPS stations. As well, a

new generation of InSAR satellites such as Sentinel, TerraSAR-X, and CosmoSkyMed is essential.

## 5. Conclusion

By comparing Mexico City's subsidence monitoring history (Table 5) with existing research and the current study, this study focused on the use of larger InSAR temporal baselines and more CGPS temporal baselines. We concluded that the combination of InSAR and CGPS temporal baselines with other available remotely sensed data utilizing SVM could improve our understanding of subsidence visualization. Coupling GPS, InSAR, and optical remotely-sensed analysis could lead to the accurate monitoring of subsidence in the area of Mexico City by exploiting the strengths and minimizing the weaknesses of each technique

The proper low spatial resolution CGPS stations (only nine stations) with east, north, and vertical components are used to calibrate the field observation assessment of the high spatial resolution PS displacement rates (more than 600,000 points). Fig. 9 shows a good correlation between CGPS and PSI data. PSI is a relative method by nature despite CGPS methodology, which provides absolute displacement rates. This is why we combined PSI with CGPS stations.

This study combined InSAR, CGPS, and optical remotely-sensed imageries to measure Mexico City's subsidence because of groundwater extraction and related risk assessments. Fifty-two ENVISAT-ASAR imageries and nine CGPS stations are used to study the deformations. Geodetic analyses based on InSAR and GPS methodologies give promising results for monitoring deformation rates in large areas. Long-term deformation rates based on PSI and GPS methodologies are similar to those found in previous InSAR and PSI works. The subsidence in the eastern part of the Mexico City metropolitan area shows fast and constant rates. This study concluded that the maximum subsidence rate is 352 mm/year in the LOS direction. This

subsidence rate is occurring in the central and eastern parts of Mexico City. The InSAR/PSI analysis results show almost perfect agreement with the GPS data at R2 in the order of 0.98 in most of the CGPS stations.

A negligible amount of observed seasonal variations in the UP component of CGPS stations and PSI time series leads to the conclusion that the recharge of aquifers in the rainy seasons is small (withdrawal over recharge). The fast subsidence rates in the Mexico City metropolitan area are a result of a massive amount of well pumping. The consequence of this is clay-rich aquifers' compactions and the permanent loss of porosity and reservoir capacity. The data we used in this study show that the mitigations have no effect on long-term compaction of the clay-rich aquifers, and the seasonal variations are small. The data also show that there is a considerable amount of risk in the metropolitan area of Mexico City. The subsidence (see [Table 5](#)) leads to the damage of buildings and infrastructures as well as economic ramifications. As the clay-rich aquifers shrink, fracturing, sinkholes, and faults develop and cause direct effects on the engineered structures.

This study used SVM classification for classifying the populated area and finding its correlation with the high PSI deformation rates. The maximum amount of subsidence is occurring in the highly populated zone of the Mexico City metropolitan area. PSI and SVM are two valuable methods to study these kinds of subsidence threats in the metropolitan areas.

Following further assessment of subsidence, mechanism of alluvial compaction, change of porosity and permeability, and geohazards, this study suggests that additional geophysical work is needed to map the exact extension of the subsidence and subsurface geology. Geodetic surveys using denser CGPS networks could estimate the precise amount of subsidence and its extensions. Regarding InSAR and PSI, using the new and improved generation of InSAR images such as

TerraSAR-X and CosmoSkyMed imageries could provide more information about the existing subsidence.

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