

Integration of Multi-Constellation GNSS Precise Point Positioning and MEMS-Based Inertial Systems Using Tightly Coupled Mechanization

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Received 21 September 2015; accepted 1 November 2015; published 4 November 2015

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Abstract

We develop a new integrated navigation system, which integrates multi-constellations GNSS precise point positioning (PPP), including GPS, GLONASS and Galileo, with low-cost micro-electromechanical sensor (MEMS) inertial system, for precise positioning applications. To integrate GNSS and the MEMS-based inertial system, the process and measurement models are developed. Tightly coupled mechanism is adopted, which is carried out in the GNSS raw measurements domain. Both un-differenced and between-satellite single-difference (BSSD) ionosphere-free linear combinations of pseudorange and carrier phase GNSS measurements are processed. Rigorous models are employed to correct GNSS errors and biases. The GNSS inter-system biases are considered as additional unknowns in the integrated error state vector. The developed stochastic model for inertial sensors errors and biases are defined based on first order Gaussian Markov process. Extended Kalman filter is developed to integrate GNSS and inertial measurements and estimate inertial measurements biases and errors. Two field experiments are executed, which represent different real-world scenarios in land-based navigation. The data are processed by using our developed Ryerson PPP GNSS/MEMS software. The results indicate that the proposed integrated system achieves decimeter to centimeter level positioning accuracy when the measurement updates from GNSS are available. During complete GNSS outages the developed integrated system continues to achieve decimeter level accuracy for up to 30 seconds while it achieves meter-level accuracy when a 60-second outage is introduced.

Keywords

GNSS, GPS, Galileo, GLONASS, MEMS, PPP, Tightly Coupled

How to cite this paper: Rabbou, M.A. and El-Rabbany, A. (2015) Integration of Multi-Constellation GNSS Precise Point Positioning and MEMS-Based Inertial Systems Using Tightly Coupled Mechanization. *Positioning*, **6**, 81-95. http://dx.doi.org/10.4236/pos.2015.64009

1. Introduction

Global navigation satellite systems (GNSS) provide worldwide positioning, velocity and time synchronization. Traditionally, highly accurate GNSS positioning solution is obtained through carrier-phase observables in differential mode involving two or more receivers. However, the requirement of a base station is usually problematic for some applications. Comparable positioning accuracy, without requiring extra infrastructure, can be achieved through precise point positioning (PPP) technique [1]. PPP uses either un-differenced or between-satellite single difference carrier-frequency and pseudorange observations from a single receiver, in addition to precise orbit and clock products. PPP commonly employs un-differenced ionosphere-free linear combination of GPS observations. Unfortunately, GPS often experiences poor satellite visibility or weak constellation geometry in urban areas. This limitation can be overcome through combining multi-constellation GNSS, which is not simply achieved by adding the additional measurements to existing GPS observation models. Inter-system biases exist, which must be taken into account in order to make effective use of the additional GNSS observation.

Employing multi-GNSS systems, in contrast to GPS only, decreases the probability of partial GNSS outages due to the availability of a large number of satellites observations. However, GNSS positioning solution may not always be available due to complete GNSS outages in urban canyons. These limitations can be overcome through integrating the GNSS observations with a relatively environment-independent system, the inertial navigation system (INS). Differential GPS are traditionally used for precise positioning applications with different grade levels of inertial sensors such as a navigation grade inertial system (e.g. [2] [3]), and a tactical grade INS (e.g. [4] [5]). Typically, previous research employed high-end INS to enhance the GPS solution. Petovello et al. (2003) [4] used high-end INS to shorten the ambiguity search time following brief GPS data outages by feeding the estimation filter with position and position variance-covariance matrix. As well, inertial sensor measurements were used to identify the GPS cycle slip, which in turn improved GPS reliability [6]. Unfortunately, highend inertial sensors are expensive and may not provide a cost effective solution. Advances in micro-electromechanical sensors (MEMS) provide the development of a generation of low-cost inertial sensors, which make them attractive to many users. However, in general, MEMS sensors have poorer performance and stability compared with high-end INS due to the high noise level and severe biases and drifts affecting the MEMS-based inertial sensors. A number of researchers have investigated the integration of GPS system with MEMS-based inertial sensors (e.g. [7]-[9]). Most of the previous research either employed the differential or classical single point positioning GPS. As such, severe positioning errors were introduced during the GPS outages, which restricted the applications of those systems. More recently, the PPP is presented in the integration system in a number of studies [10]-[18]. However, these studies were based on the pseudoorange and carrier phase observations of a single GNSS constellation, namely GPS.

Considering the recent advances in MEMS-based accelerometers, the up to date GNSS constellations and the advances in PPP techniques, this research aims to develop a new integrated navigation system for precise positioning and navigation applications. MEMS-based accelerometers equipped with fiber optic gyros, which limit the orientation errors, are used. GNSS-based PPP including GPS, GLONASS and Galileo systems observations are used to update the system through a tightly coupled mechanism. The developed integrated system shows decimetre to centimetre level accuracy when GNSS observations are available. It is shown that the additional GNSS observations enhance the positioning accuracy in comparison with the traditional GPS kinematic positioning solution. Better positioning accuracy is obtained with BSSD ionosphere-free model, in comparison with the traditional un-differenced ionosphere-free model. In addition, the developed integrated system continues to achieve decimeter level accuracy for up to 30 seconds while it achieves meter-level accuracy when a 60-second outage is introduced.

2. Multi-Constellation GNSS-PPP Measurement Models

In this study, both un-differenced and between-satellite single differenced ionosphere-free models are considered. Pseudorange and carrier phase observations of three GNSS systems are processed, namely GPS, GLONASS and Galileo. The general un-differenced ionosphere-free linear combinations of GNSS observations can be written as [19]:

$$P_{3} = \rho + c \left[dt_{r} + B^{r} \right] - c \left[dt^{s} - B^{s} \right] + T + c \left[ISB \right] + e$$

$$\tag{1}$$

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$$\Phi_{3} = \rho + c \left[dt_{r} + B^{r} \right] - c \left[dt^{s} - B^{s} \right] + T + c \left[ISB \right] + \left(\overline{\lambda} \, \overline{N} + \Delta B^{r} - \Delta B^{s} \right) + \varepsilon$$
⁽²⁾

where B^r , B^s are ionosphere-free differential code biases for receiver and satellites, respectively; *ISB* is the inter-system bias which is the difference between receiver differential code bias of the GPS and the other GNSS systems. The inter-system bias for GPS is equal zero; ΔB^r is the difference between receiver differential code and phase biases; ΔB^s is the difference between satellite differential code and phase biases. As can be seen from Equation (1) to Equation (2), the un-calibrated biases such as ΔB^r and ΔB^s are lumped with the GNSS ambiguity parameters.

The IGS-MGEX precise orbital and clock products are used to mitigate the satellite orbit and clock errors [20]. The UNB3 tropospheric model, consisting of the Saastamoinen vertical propagation delay model and Niell mapping function, is used to account for the dry tropospheric component [21]. The effects of ocean loading, Earth tide, carrier-phase windup, sagnac, relativity, and satellite antenna phase-center variations are rigorously modeled as detailed in [22]. As a result, the mathematical model for the un-differenced GNSS ionosphere-free observations can be simplified to

$$\overline{P}_{3} = \rho + cd\overline{t_{r}} + m^{s}T_{w} + c[ISB] + e$$
(3)

$$\overline{\Phi}_{3} = \rho + c\overline{d}t_{r} + m^{s}T_{w} + c[ISB] + B + \varepsilon$$
(4)

where \overline{P}_3 and $\overline{\Phi}_3$ are the corrected pseudorange and carrier phase measurements, respectively \overline{dt}_r is receiver clock error lumped with GNSS receiver differential code bias; $m^s = \frac{1}{\sin(\text{elevation})}$ is the mapping

function for the troposphere wet delay component T_w ; *B* is the float ambiguity in meters as described in Equation (2). To develop the mathematical equations for BSSD, we refer to the GNSS satellite by *k*. GPS satellite *l* is taken as the reference satellite to form tight BSSD ionosphere-free linear combination.

$$\overline{P}_{3}^{kl} = \rho^{kl} + m^{kl}T_{w} + c[ISB] + e^{kl}$$

$$\tag{5}$$

$$\overline{\Phi}_{3}^{kl} = \rho^{kl} + m^{kl}T_{w} + c[ISB] + B^{kl} + \varepsilon^{kl}$$
(6)

It can be seen that the receiver clock offset is cancelled out when forming our BSSD mathematical equations. Additionally, the receiver differential code and phase biases are cancelled out for the GPS system observations while these receiver biases are reduced significantly for GLONASS and Galileo observations. However, forming BSSD leads to mathematical correlations among the observations, which must be taken into account when the covariance matrix of the observations is formed. Equations (3)-(6) are used to develop the measurement models of the proposed GNSS/INS integrated system for both un-differenced and between satellites single differences modes, respectively. However, due to the nonlinearity of GNSS observation models, the GNSS mathematical model should be expanded through Tylor Expansion to be employed in updating the tight PPP/INS integration as follows

For undifferenced GNSS ionsphere-free model;

$$\overline{P}_{3} - \overline{P}_{INS}^{0} = -D\delta r + c\overline{d}t_{r} + m^{s}T_{w} + cISB + e$$

$$\tag{7}$$

$$\overline{\Phi}_{3} - \Phi^{0}_{INS} = -D\delta r + c\overline{d}t_{r} + m^{s}T_{w} + B + cISB + \varepsilon$$
(8)

And for BSSD GNSS ionosphere-free model:

$$\overline{P}_{3}^{kl} - \overline{P}_{INS}^{kl} = -D^{kl}\delta r + m^{kl}T_{w} + cISB + e^{kl}$$
⁽⁹⁾

$$\overline{\Phi}_{3}^{kl} - \Phi_{INS}^{kl} = -D^{kl}\delta r + m^{kl}T_{w} + cISB + B^{kl} + \varepsilon^{kl}$$

$$\tag{10}$$

where P_{INS} and Φ_{INS} are the predicted INS pseudorange and carrier phase measurements. *D* is the direction cosine vector from the receiver to the satellite: δr is a three-dimensional vector representing the positioning errors.

2.1. Inertial Navigation Motion Model

Inertial navigation is a method where the current position, velocity and attitude of a moving object are deter-

mined from a history of acceleration and angular velocity measurements. Acceleration and angular velocity are measured using accelerometers and gyros. Unlike GNSS systems, the INS performance is not affected in environments as urban canyons; it is independent of external electro-magnetic signals. However, the main drawback of an INS is the degradation of its performance with time. In order to control the errors to an acceptable level continues updates from, for example, GNSS are necessary.

The mathematical model of the inertial navigation system is commonly described in the framework of linear dynamic systems. The dynamic behavior of such systems can be described by using a state-space representation. For this purpose, a system of non-linear first-order differential equations can be described as [23]:

$$\begin{bmatrix} \dot{r}^{e} \\ \dot{V}^{n} \\ \dot{R}^{n}_{b} \end{bmatrix} = \begin{bmatrix} C \cdot V^{n} \\ R^{n}_{b} f^{b} - (2\Omega^{n}_{ie} + \Omega^{n}_{en}) \cdot V^{n} + g^{n} \\ R^{n}_{b} (\Omega^{n}_{ib} - \Omega^{b}_{in}) \end{bmatrix}$$
(11)

where r^e is the position vector, latitude, longitude and altitude; *C* is a transformation matrix from the East, North and Up (ENU) reference frame to earth-centered earth-fixed (ECEF) frame (Jakili, 2001); V^n is the velocity vector in ENU reference frame, \dot{V}^n is the kinematic acceleration vector in the ENU reference frame; $\Omega^n_{en} \cdot V^n$ represents the effect of the motion of the ENU frame with respect to the ECEF frame; $2\Omega^n_{ie} \cdot V^n$ is the Coriolis acceleration vector; g^n is the gravity vector, including the gravitation term and the centripetal term related to the Earth rotation; and f^b is the specific force vector in the body frame, which is measured by the accelerometers. The matrix Ω^n_{ie} is the skew-symmetric matrix of rotation rate vector of the Earth, which can be expressed in the ENU frame as:

$$\Omega_{ie}^{n} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & \omega \cos \varphi & 0 \\ 0 & 0 & \omega \sin \varphi \end{bmatrix}$$
(12)

The matrix Ω_{en}^{n} is the skew-symmetric matrix of the rotation rate vector of the ENU frame with respect to ECEF frame, expressed in the ENU frame as:

$$\Omega_{en}^{n} = \begin{bmatrix} \frac{-V_{n}}{M+h} & 0 & 0\\ 0 & \frac{V_{E}}{N+h} & 0\\ 0 & 0 & \frac{V_{E}\tan\varphi}{N+h} \end{bmatrix}$$
(13)

The matrix Ω_{ib}^{b} is the skew-symmetric matrix of the rotation rate vector of the body frame with respect to the ECI frame ω_{ib}^{b} , expressed in the body reference, which is measured by the gyros. The matrix Ω_{in}^{b} is the skew-symmetric matrix of the rotation rate of the navigation frame with respect to inertial frame ω_{in}^{b} expressed in the body frame, which is computed combining ω_{ie}^{n} and ω_{en}^{n} transforming the result in the body frame as follows:

$$\Omega_{in}^{b} = R_{n}^{b} \cdot \left(\Omega_{ie}^{n} + \Omega_{en}^{n}\right) \tag{14}$$

2.2. GNSS-PPP/MEMS-Based IMU Implementation

To build the proposed GNSS/INS integrated navigation system, tightly coupled architecture is implemented adopting extended Kalman filter (EKF). GNSS pseudorange, carrier phase and Doppler measurements as well as INS-derived observations are processed to produce estimates of the state vector including position, velocity and attitude. The precise GNSS ephemerides as well as the outputs of position P^n and velocity V^n from the inertial sensors mechanization are used to predict the INS pseudorange P_{INS} , phase Φ_{INS} and Doppler \dot{P}_{INS} measurements. The corrected GNSS pseudorange P_{GNSS} , phase Φ_{GNSS} and Doppler \dot{P}_{GNSS} measurements are differenced with the INS-predicted measurements. The residuals δP , $\delta \Phi$ and $\delta \dot{P}$ are then directly processed

by the integration filter to estimate the system error state vector. The obtained INS error estimates, such as the inertial sensors bias drifts δb_a and δb_g , and scale factors δS_a and δS_g , are fed back to the INS mechanization to correct for the inertial sensors forces f^b and w^b using the closed loop approach. The estimated error states such as position errors δr^n , velocity errors δv^n and attitude errors $\delta \varepsilon^a$ are directly applied to the INS-derived position P_{INS}^n , velocity V_{INS}^n and attitude A_{INS}^n solutions. States unique to GNSS such as the clock offset δt_{off} , clock drift δt_{dri} , internal system biases *ISB* and ambiguity parameters *N*, are fed back to continue correct for the GNSS observations using additional closed loop technique. A priori estimation constrains are applied on GPS/GLONASS and GPS/Galileo internal system biases (ISBs) to continue benefits from additional GNSS satellites when the number of GLONASS or Galileo satellites drops to one satellite. Figure 1 shows the tightly coupled GNSS PPP/INS implementation flowchart.

To implement the mechanization of the developed integrated system, the EKF is used as an estimator to merge the GNSS observations and INS records. The estimated state vector δx consists of 26 + n states describing the basic state vector including the nine navigation parameter errors, the inertial sensors errors including the bias drift and scale factor, and errors unique to the GNSS measurements, which are mainly the receiver clock offset and drift, the troposphere wet delay component, the GPS/GLONASS ISB and GPS/Galileo ISB with additional n states related to the float ambiguity parameters Bi. The complete state vector for un-differenced ionosphere-free technique can be written as.

$$\delta x = \left[\delta r^{n}, \delta v^{n}, \delta \varepsilon^{n}, \delta b_{a}, \delta b_{g}, \delta S_{a}, \delta S_{g}, T_{w}, c d t_{r}, c \delta t_{dri}, c I S B_{R}, c I S B_{E}, B_{1}, B_{2}, \cdots, B_{n} \right]$$
(15)

where δr^n is a three-dimensional vector representing the positioning errors in latitude, longitude and altitude, δv^n is a three-dimensional vector representing thevelocity errors in east, north and up, $\delta \varepsilon^n$ is a three-dimensional vector representing the attitude errors in roll, pitch and azimuth, δb_a is a three-dimensional vector representing theaccelerometer biases drift in x, y and z, δb_g is a three-dimensional vector representing the gyro biases drift in x, y and z, δS_a is a three-dimensional vector representing the accelerometer scale factors errors in x, y and z, δS_g is a three-dimensional vector representing the gyro scale factors errors in x, y and z. T_w is the wet component of the tropospheric delay, \overline{dt}_r and $c\delta t_{dri}$ are the GPS receiver clock offset and drift in

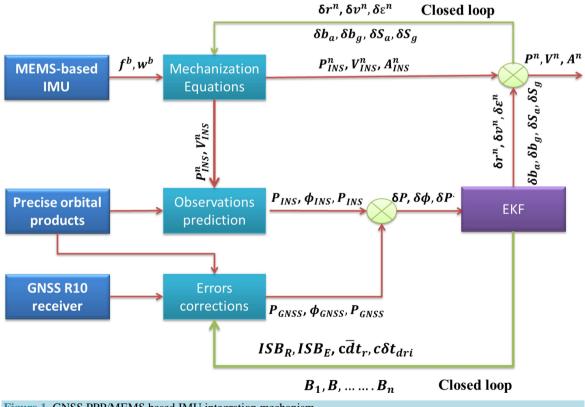


Figure 1. GNSS PPP/MEMS based IMU integration mechanism.

meters, respectively. Both $cISB_R$ and $cISB_E$ are GPS/GLONASS and GPS/Galileo inter-system biases in meters. *B* is the float ambiguity term in meters. It should be pointed out that the receiver clock offset and drift are cancelled out when forming BSSD ionosphere-free model.

EKF includes two parts the system model and the observation model. The system model is obtained from the INS dynamic errors augmented with the additional GNSS errors as follows.

$\delta \dot{r}^n$		03		(7		0	3		03	03		03		0	1	0	0	0	0	0	0	0	0	δr^n	
$\delta \dot{v}^n$		0_{3}^{2}	_	$2\Omega_{ie}^n$	$-\Omega_e^n$	n	R_b^n	-		\mathbf{R}_{b}^{n}	0,		$R_b^n F$	b	0		0	0	0	0	0	0	0	0	δv^n	
$\dot{arepsilon}^n$		03		0			2^{n}_{ib} -			0,	R_b'		03		$R_b^n V$		0	0	0	0	0	0	0	0	ε^{n}	
$\delta \dot{b}_a$		03		0	3		0	3	ļ	B_{ba}	0,		03		0	3	0	0	0	0	0	0	0	0	δb_a	
$\delta \dot{b}_{g}$		03		0	3		0			03	β_{b}		03		0	3	0	0	0	0	0	0	0	0	δb_{g}	
$\delta \dot{S}_a$		03		0	3		0	3	(03	03		$\beta_{\scriptscriptstyle sa}$		0	3	0	0	0	0	0	0	0	0	δS_a	
$\delta \dot{S}_{g}$		03		0	3		0	3	(03	03		03		β_{s}	a	0	0	0	0	0	0	0	0	δS_{g}	
$\delta \dot{T}_{_{W}}$	=	0		()		C)		0	0		0		0		0	0	0	0	0	0	0	0	δT_w	
$\dot{\delta}cdt_r$		0		()		C)		0	0		0		0		0	1	1	0	0	0	0	0	δcdt_r	
$\dot{\delta}ct_{dri}$		0		()		C)		0	0		0		0		0	0	0	0	0	0	0	0	δct_{dri}	
$\delta cISB_{R}$		0		()		C)		0	0		0		0		0	0	0	0	0	0	0		$\delta cISB_{R}$	
$\delta cISB_{E}$		0		()		C)		0	0		0		0		0	0	0	0	0	0	0	0	$\delta cISB_{E}$	
$\delta \dot{B}_{G}$		0		()		C)		0	0		0		0		0	0	0	0	0	0	0	0	δB_{G}	
$\delta \dot{B}_{R}$		0		()		C)		0	0		0		0		0	0	0	0	0	0	0	0	δB_R	
$\delta \dot{B}_{E}$		0		()		C)		0	0		0		0		0	0	0	0	0	0	0	0	δB_{E}	
			03	03	03	03	03	03	0	0	0	0	0	0	0	0	W _a	I								
			R_b^n	03	03	03	03	03	0	0	0	0	0	0	0	0	w_{g}									
			03	R_b^n	03	03	03	03	0	0	0	0	0	0	0	0	W_{ba}									
			03	03	I_3	03	03	03	0	0	0	0	0	0	0	0	w_{bg}									
			03	03	03	I_3	03	03	0	0	0	0	0	0	0	0	W_{sa}									
			03	03	03	03	I_3	03	0	0	0	0	0	0	0	0	W_{sg}									
			03	03	03	03	03	I_3	0	0	0	0	0	0	0	0	W _{tw}									
			03	03	03	03	03	03	1	0	0	0	0	0	0	0	W_{cdt_r}									
			03	03	03	03	03	03	0	1	0	0	0	0	0	0	w _{dri}									(1c)
			03	03	03	03	03	03	0	0	1	0	0	0	0	0	w _{ISB}									(16)
			03	03	03	03	03	03	0	0	0	1	0	0	0	0	W _{ISB}									
			03	03	03	03	03	03	0	0	0	0	1	0	0	0	0									
			03	03	03	03	03	03	0	0	0	0	0	0	0	0	0									
			03	03	03	03	03	03	0	0	0	0	0	0	0	0	0									
			03	03	03	03	03	03	0	0	0	0	0	0	0	0	I	1								

where R_b^n is the transformation matrix from the body frame to the navigation frame, F^b is a diagonal matrix of the accelerometers forces in body frame and W^b Is a diagonal matrix of the gyro forces in body frame, w representing the system input white noise, G is the associated coefficient matrix and, $\beta = 1/\tau$, where τ is the correlation time for the accelerometers and gyros for first order GM process. The approximate correlation times are estimated by computing the autocorrelation functions for accelerometer and gyros records using static data records collected for three hours [5]. The observation model of the GNSS/INS filter in the tightly coupled architecture has the typical form:

$$\delta Z = H \cdot \delta x + \eta \tag{17}$$

 δZ is the measurement vector consisting of the differences between the corrected GNSS and the predicted INS measurements. When un-differenced ionosphere-free model is used δZ can be defined as:

$$\delta Z = \begin{bmatrix} P_i^G - P_i^{ins} \\ P_h^R - P_h^{ins} \\ P_n^E - P_n^{ins} \\ \vdots \\ \phi_i^G - \phi_i^{ins} \\ \phi_h^R - \phi_h^{ins} \\ \phi_n^E - \phi_n^{ins} \\ \vdots \\ \dot{P}_i^G - \dot{P}_i^{ins} \\ \dot{P}_h^R - \dot{P}_h^{ins} \\ \dot{P}_h^R - \dot{P}_h^{ins} \\ \dot{P}_n^E - \dot{P}_n^{ins} \\ \vdots \end{bmatrix}$$
(18)

H is the design matrix containing geometry factors defined according to the GNSS mathematical model used. The design matrix is arranged with columns corresponding to the states unique to inertial sensors errors such as δb_a , δb_g , δS_a and δS_g which filled with zeroes. *H* can be formed as:

where d are the direction cosine matrix D elements for pseudorange and phase; s is the direction cosine matrix S elements for Doppler measurements. The Element of D and S can be computed as follows

$$D = \frac{\begin{pmatrix} x_{ins} - X^G \\ \rho_{ins} \end{pmatrix}}{\begin{pmatrix} x_{ins} - X^R \\ \rho_{ins} \end{pmatrix}} \frac{y_{ins} - Y^G \\ \rho_{ins} \end{pmatrix}}{\begin{pmatrix} x_{ins} - X^R \\ \rho_{ins} \end{pmatrix}} \frac{y_{ins} - Y^R \\ \rho_{ins} \end{pmatrix}}{\begin{pmatrix} x_{ins} - X^E \\ \rho_{ins} \end{pmatrix}} \frac{y_{ins} - Y^E \\ \rho_{ins} \end{pmatrix}}{\begin{pmatrix} y_{ins} - Y^E \\ \rho_{ins} \end{pmatrix}} \frac{y_{ins} - Z^E \\ \rho_{ins} \end{pmatrix}}{\begin{pmatrix} -(N+h)\cos\lambda\sin\varphi & -(N+h)\sin\lambda\cos\varphi & \cos\lambda\cos\varphi \\ -(N+h)\sin\lambda\sin\varphi & -(N+h)\cos\lambda\cos\varphi & \sin\lambda\cos\varphi \\ (N(1-e^2)+h)\cos\varphi & 0 & \sin\varphi \\ \end{pmatrix}$$
(20)
$$\begin{pmatrix} x_{ins} - X^R \\ p_{ins} \end{pmatrix} \frac{y_{ins} - Y^R \\ p_{ins} \end{pmatrix} \frac{z_{ins} - Z^R }{\rho_{ins}} \frac{y_{ins} - Z^R }{\rho_{ins}} \\ \frac{x_{ins} - X^R }{\rho_{ins}} \frac{y_{ins} - Y^R }{\rho_{ins}} \frac{z_{ins} - Z^R }{\rho_{ins}} \\ \vdots & \vdots & \vdots \\ \end{pmatrix}$$

$$S = \begin{vmatrix} \frac{x_{ins} - X^{G}}{\rho_{ins}} & \frac{y_{ins} - Y^{G}}{\rho_{ins}} & \frac{z_{ins} - Z^{G}}{\rho_{ins}} \\ \frac{x_{ins} - X^{R}}{\rho_{ins}} & \frac{y_{ins} - Y^{R}}{\rho_{ins}} & \frac{z_{ins} - Z^{R}}{\rho_{ins}} \\ \frac{x_{ins} - X^{E}}{\rho_{ins}} & \frac{y_{ins} - Y^{E}}{\rho_{ins}} & \frac{y_{ins} - Z^{E}}{\rho_{ins}} \\ \vdots & \vdots & \vdots & \vdots \end{vmatrix} \begin{vmatrix} -\sin\lambda & -\cos\lambda\sin\varphi & \cos\lambda\cos\varphi \\ \cos\lambda & -\sin\lambda\sin\varphi & \sin\lambda\cos\varphi \\ 0 & \cos\varphi & \sin\varphi \end{vmatrix}$$
(21)

where *X*, *Y* and *Z* are are the satellite coordinates computed using the final IGS-MEGX orbital products and corrected for the effect of earth rotation during signal transit; φ , λ and *h* are the INS positioning coordinates; *N* is the prime vertical radius of curvature. To form the BSSD measurement model, between-satellite single difference matrix M_{bssd} should be defined based on the selected GPS reference satellite.

$$M_{bssd} = \begin{vmatrix} -1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 & 1 & \vdots & \vdots & \vdots & \vdots \\ \vdots & 0 & \ddots & 0 & 1 & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & 0 & -1 & 1 & 0 & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & 0 & 1 & -1 & 0 & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & 1 & 0 & \ddots & 0 & \vdots \\ \vdots & \vdots & \vdots & \vdots & 1 & 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & -1 \end{vmatrix}$$

$$H_{BSSD} = M_{bssd} * H_{un-diff}$$

$$(22)$$

$$\delta Z_{bssd} = M_{bssd} * \delta Z_{un-diff} \tag{24}$$

where H_{bssd} is the design matrix for BSSD model, δZ_{bssd} is the BSSD observation vector. The error state vector for BSSD based integrated system is defined as:

$$\delta x = \left[\delta r^{n}, \delta v^{n}, \delta \varepsilon^{n}, \delta b_{a}, \delta b_{g}, \delta S_{a}, \delta S_{g}, \Delta T_{wli}, cISB_{R}, cISB_{E}, B_{li}, B_{2i}, \cdots\right]$$
(25)

where B_{1i} and B_{2i} are the single differenced float ambiguity terms.

2.3. Tests and Results Analysis

Two real vehicular tests were conducted to evaluate the performance of the developed integrated GNSS-PPP/ MEMS-based INS system (Figure 2). The vehicular tests were carried out through downtown Kingston, Ontario, Canada, which was designed to represent challenging situations for real GNSS satellite navigation availability including turns, straight portions, high speed, and slow speeds. The NovAtel SPAN-CPT system and the GNSS Trimble R10 receiver were used to collect the navigation data. The SPAN-CPT system consists of the NovAtel OEM4 receiver and a MEMS-based IMU, which contains three MEMS-based accelerometers and three fiber optic gyros. Carrier phase-based differential GNSS (DGNSS) solution is used as a reference solution. In order to create this reference solution, a GNSS Trimble R7 receiver was setup at a nearby station with precisely known coordinates. The raw dual-frequency GNSS pseudorange, carrier phase and Doppler measurements were collected at a 1 Hz rate, while the IMU raw data was logged at a rate of 100 Hz. The duration of the first trajectory test was set for about 55 minutes while the duration of the second test was 34 minutes. Four scenarios are considered in this research. The traditional GPS-based integrated system and the developed GNSS-based integrated system including GPS, GLONASS and Galileo, are studied to investigate the contribution of the additional GNSS systems to the positioning accuracy of the integrated system. Both un-differenced and BSSD ionosphere-free PPP techniques are adopted for GPS-based and GNSS-based integrated systems. To investigate the positioning accuracy of the integrated system during complete GNSS outages, a number of simulated outages is

introduced for each trajectory test. The data were processed using our Ryerson PPP GNSS/MEMS software in un-difference and BSSD modes. The program is implemented in MATLAB R2013a using the Intel[®] Core i7-3517U CPU and 6 GB RAM. The computational burden of the whole process is 39.41 s including reading both INS and GNSS observations, Kalman filtering process with GNSS updating every second and results writing.

2.3.1. First Trajectory

The first trajectory test area is shown in **Figure 3** with the locations of simulated outages. **Figure 4** shows the GNSS satellite availability during the observation time.





Figure 2. Equipment setup.



Figure 3. Test area and simulated complete GNSS outages for the first trajectory.

Figure 5 shows the positioning accuracy of the developed integrated system when the observations of all GNSS satellites are included in the solution, *i.e.*, no outages are inserted. It can be seen that the addition of GLONASS and Galileo observations enhances the positioning accuracy and convergence time in comparison with the GPS-only positioning solution. Further improvement is attained in the positioning solution through BSSD ionosphere-free linear combination model, in comparison with the traditional un-differenced counterpart.

Table 1 summarizes the statistical characteristics, mainly the root mean square error (RMSE) and the maximum error after the convergence time, for the four PPP integrated system scenarios mentioned above. Comparing the RMSE values for each scenario, it can be seen that the positioning precision is improved by 40%, 41% and 41% for latitude, longitude and altitude in the multi-constellation GNSS PPP solution compared with the GPS-only PPP solution. In addition, using BSSD ionosphere-free PPP technique improves the positioning precision case by 23%, 15% and 13% for latitude, longitude and altitude, in comparison with the traditional un-differenced ionosphere-free PPP technique.

To mimic challenging positioning conditions in urban areas, including complete blockage of the GNSS satellites, twelve simulated complete satellite outages of 60 s, 30 s and 10 s were introduced in the first trajectory.

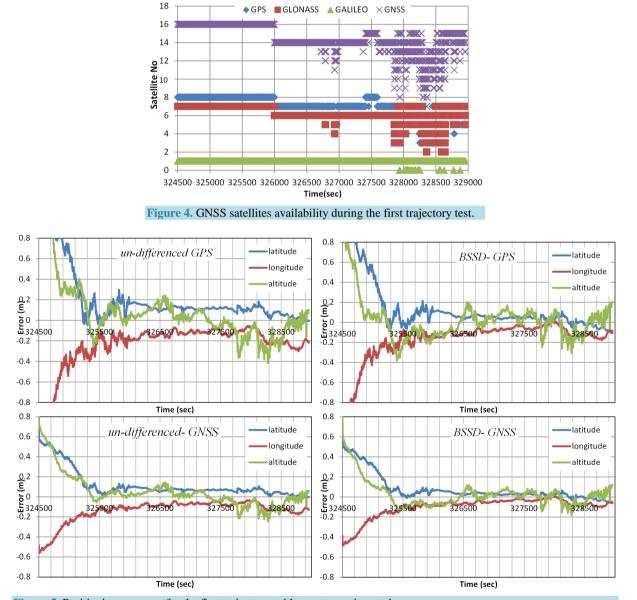
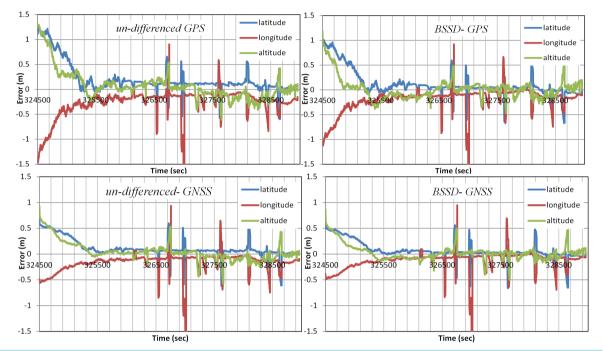


Figure 5. Positioning accuracy for the first trajectory, with no outages inserted.

Figure 6 shows the positioning errors during the various outages, referenced to carrier-based DGPS solution with full satellite availability. As can be seen, both of the un-difference and BSSD models produce similar positioning accuracy during the outages. In addition the contribution of the additional GNSS systems observation can be considered marginal, as the positioning error during a complete GNSS outage depends on the accuracy of the positioning solution just before the occurrence of outage. As well, the additional GNSS observations can only slightly improve the inertial sensor bias estimation, compared with that of GPS-only. In the 60-second GNSS outage the maximum positioning error reached meter level in most cases, while it reached a decimeter level in 10-seccond outage. Table 2 shows the average maximum positioning errors in latitude, longitude and





PPP techniques	GPS	8 (un-differenced m	ode)		GPS (BSSD mode)			
Positioning	latitude	longitude	altitude	Latitude	longitude	altitude		
RMSE (m)	0.101	0.160	0.103	0.052	0.090	0.082		
Maximum error	0.184	0.303	0.416	0.121	0.179	0.306		
PPP techniques	GNS	S (un-differenced n	node)	GNSS (BSSD mode)				
Positioning	latitude	longitude	altitude	Latitude	longitude	altitude		
RMSE	0.065	0.094	0.079	0.034	0.059	0.058		
Maximum error	0.108	0.178	0.245	0.072	0.106	0.180		

Table 1.	Statistical	l analysis of	GNSS	positioning	precision f	or the f	irst trajectory	, with	no outages inserted.
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Table 2. Average maximum positioning errors during GNSS simulated outages for the first trajectory.

PPP technique	U	n-differenced-GF	PS	BSSD-GPS				
Outages(sec)	60 s	30 s	10 s	60 s	30 s	10 s		
Latitude(m)	0.517	0.334	0.201	0.501	0.327	0.199		
Longitude(m)	0.716	0.429	0.214	0.699	0.428	0.210		
Altitude(m)	0.402	0.310	0.159	0.393	0.305	0.160		
PPP technique	Uı	n-differenced-GN	SS	BSSD-GNSS				
Outages(sec)	60 s	30 s	10 s	60 s	30 s	10 s		
Latitude(m)	0.483	0.296	0.175	0.472	0.268	0.146		
Longitude(m)	0.681	0.396	0.186	0.670	0.363	0.159		
Altitude(m)	0.376	0.273	0.137	0.357	0.245	0.104		

altitude, respectively, during the three simulated GNSS outages for both BSSD and un-differenced ionospherefree models for the first trajectory.

2.3.2. Second Trajectory

The second trajectory test area is shown in **Figure 7** with the locations of simulated outages. Similar to the first trajectory, the locations of the simulated outages were selected to present different driving conditions. The second trajectory is featured with higher vehicle velocities compared with the first trajectory. **Figure 8** shows the satellite availability during the observation time.

Figure 9 shows the positioning accuracy for the developed integrated system when the observations of all GNSS satellites are included in the solution, *i.e.*, no outages are inserted. **Table 3** summaries the statistical analysis for the results of the four scenarios, as described earlier. It can be seen that the solution characteristics of the second trajectory are similar to those of the first trajectory, which confirms the consistency of the positioning solution.

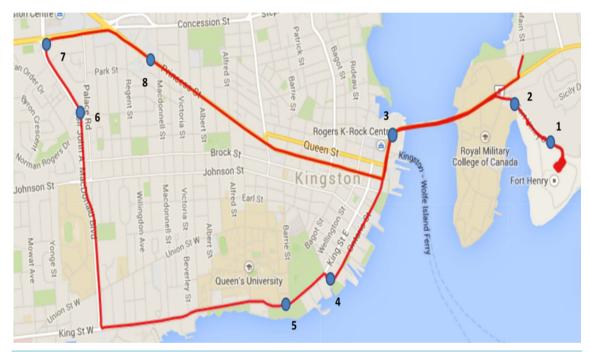


Figure 7. Test area and simulated complete GNSS outages for the second trajectory.

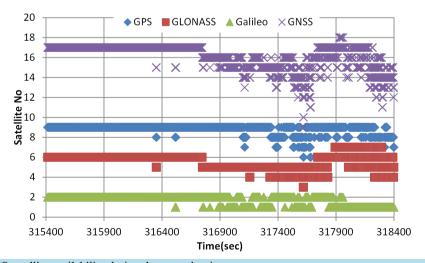


Figure 8. GNSS satellite availability during the second trajectory test.

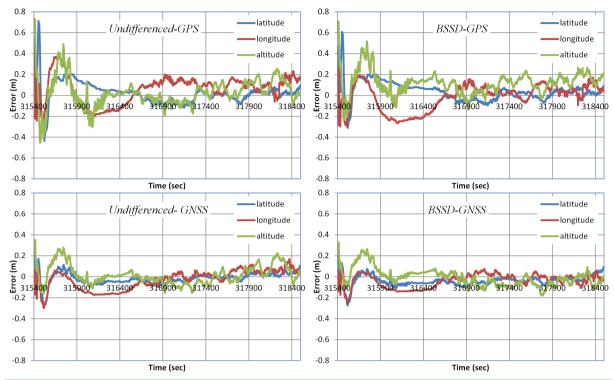


Figure 9. Positioning accuracy for the second trajectory, with no outages inserted.

Table 3. Statistical analysis of GNSS p			
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PPP techniques	GPS	GPS (un-differenced mode) GPS (BSS						
Positioning	Latitude	Longitude	Altitude	Latitude	Longitude	Altitude		
RMSE (m)	0.042	0.103	0.117	0.040	0.059	0.074		
Maximum error	0.118	0.232	0.268	0.123	0.172	0.255		
PPP techniques	GNS	S (un-differenced m	iode)	GNSS (BSSD mode)				
POSITIONING	Latitude	Longitude	Altitude	Latitude	Longitude	Altitude		
RMSE	0.029	0.051	0.062	0.038	0.030	0.057		
Maximum error	0.109	0.170	0.226	0.095	0.077	0.093		

Eight simulated GNSS outages, each with duration of 60 s, 30 s and 10 s, respectively, were introduced such that they encompass all conditions of the trajectory, including straight portions and turns. **Figure 10** shows the positioning errors during the GNSS simulated outages, which presents comparable positioning accuracy with the results of the first trajectory.

Table 4 shows the average maximum positioning errors in latitude, longitude and altitude, respectively during the three simulated GNSS outages for both BSSD and un-differenced ionosphere-free models for the second trajectory. Similar to those of the first trajectory, the average maximum positioning error reached meter level during the 60-second GNSS outage, while it reached a decimeter level in 10-seccond outage.

3. Conclusion

We developed new algorithms for the integration of multi-constellation GNSS PPP, including GPS, GLONASS and Galileo systems, and MEMS-based inertial system. Both un-differenced and between-satellite single difference ionosphere-free linear combinations of carrier phase and code GNSS measurements were considered. Tightly coupled mechanism was implemented and extended Kalman filter (EKF) technique was developed to merge the GNSS and inertial measurements. The performance of the newly developed models was analyzed by using two real trajectory tests. The positioning results of the integrated system showed that centimeter to

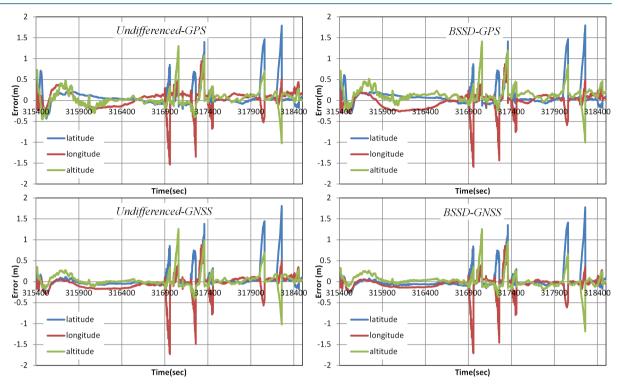


Figure 10. Positioning accuracy for the second trajectory, with simulated complete GNSS outages inserted.

PPP technique	U	n-differenced-GF	PS	BSSD-GPS				
Outages (sec)	60 s	30 s	10 s	60 s	30 s	10 s		
Latitude (m)	1.123	0.587	0.261	1.025	0.550	0.238		
Longitude (m)	1.231	0.644	0.288	1.136	0.636	0.263		
Altitude (m)	0.923	0.483	0.215	0.843	0.441	0.197		
PPP technique	Uı	n-differenced-GN	SS	BSSD-GNSS				
Outages (sec)	60 s	30 s	10 s	60 s	30 s	10 s		
Latitude (m)	1.012	0.528	0.235	0.935	0.497	0.215		
Longitude (m)	1.108	0.581	0.262	1.025	0.574	0.237		
Altitude (m)	0.832	0.441	0.194	0.769	0.398	0.180		

Table 4. Average maximum positioning errors during GNSS simulated outages for the second trajectory.

decimeter-level accuracy was achievable when the GNSS satellite were available. The addition of GLONASS and Galileo observations enhanced the positioning accuracy in comparison with standalone GPS-based solution. Better positioning accuracy was obtained with BSSD IF model in comparison with the un-differenced IF model for both GPS- and GNSS-based models. During the GNSS outages, the integrated system showed meter-level accuracy in most cases when a 60-second outage was introduced. However, the positioning accuracy was improved to a few decimeter and decimeter-level accuracy when 30- and 10-second GPS outages were introduced. Comparable results were obtained from both BSSD and un-differenced models under GNSS outages.

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