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# Differential GPS for small UAS using consumer-grade single-frequency receivers

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#### Abstract

Consumer-grade single-frequency GPS receivers with their known limitations are the predominant means of localization for small Unmanned Aircraft Systems (UAS). More intricate maneuvers such as automatic landings require a level of accuracy this class of receivers does not provide. As a contribution to improve the positioning accuracy without sacrificing the low-cost approach of this class of vehicles, a Local Area Augmentation System (LAAS) has been developed, based on consumer-grade single-frequency miniature GPS receivers both for the base station and airborne positioning.

On the part of the airborne receiver, the conventional approach of carrier phase smoothing has been extended by incorporating Doppler measurements to propagate the position during carrier phase signal outages or in the event of cycle slips. Pseudoranges and the augmented carrier phase observations are merged by means of an indirect linear Kalman filter in the position domain. The characteristics of the error state allow for some simplifications that reduce the computing effort of the filter.

To evaluate the system's performance under dynamic conditions, raw GPS data have been collected on a ground based moving platform and processed with Simulink. The results show a significantly improved 3D position accuracy compared to the standalone receiver solution.

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## 1 Introduction

Progress in low-cost sensor and computing technology within the last years has led to a growing number of new low-cost UAS designs for civil applications. Typically, they rely on single-frequency GPS receivers for localization due to their low-cost profile. Despite the efforts of manufacturers to improve the accuracy of single-frequency receivers by better modeling of atmospheric effects and the availability of space based augmentation systems such as EGNOS, today's low cost receivers are limited to 3D positioning errors of the order of several meters  $(1 \sigma)$ . This is adequate for cruise flight navigation but is expected to be a severe constraint for enlarging the flight envelope of small UAS including more advanced maneuvers such as automatic takeoffs and landings in confined areas and formation flight of multiple UAS.

In civil and military aviation, Ground Based Augmentation Systems (GBAS) are used to enhance the accuracy of GNSS positioning to a level that, for instance, allows fully automatic approaches. Unfortunately, commercial differential GNSS receivers do not meet the cost requirements of small UAS for civil applications. There are however low-cost receivers on the market that output raw code, carrier phase and Doppler observations and are thus in principle suitable for differential positioning techniques. This has been the motivation to investigate the achievable accuracy of a ground based augmentation system for UAS based on low-cost consumer-grade receivers.

There have been efforts before to use inexpensive single-frequency receivers for kinematic precise positioning applications, mostly for positioning of ground vehicles, see e.g. [4], [7]. The authors of [9] use L1 phase observations collected with a u-blox LEA-4T for the reconstruction of relative flight trajectories and obtain positive results using time-differenced carrier phase observations.

Precise absolute positioning with low cost single-frequency receivers is challenging compared to dualfrequency navigation grade receivers that are routinely used for surveying applications. Their observations typically exhibit a higher noise level, degrading the accuracy of the code based position and thus enlarging the search space for carrier ambiguity resolution. Since the ionospheric range error cannot be eliminated by forming the ionosphere-free combination of the two signals, corrections are valid only for up to about 10 km. In addition, they lack the advantage of accelerated ambiguity resolution by means of the wide-lane combination of the two carrier phase observations.

Embedded systems on board of small UAS are typically based on sub - GHz microcontrollers due to power and size constraints, thus a positioning system that requires low computing power is desirable. Moreover, continuous availability of a position estimate is of the essence to ensure operational safety. This article describes an approach towards meeting these two requirements. Results are given of a first implementation of the augmentation system in Matlab/Simulink. It has been applied to raw data collected on a ground based moving platform.

The article is organized as follows: after introducing the observation model in section 2, section 3 presents the computation of broadcast corrections and the method that is proposed to obtain the reference station position, section 4 is devoted to the positioning algorithm of the airborne receiver. Finally, section 5 presents our experimental results and section 6 gives a short summary and an outlook on future work.

# 2 Observations of a GNSS receiver

A GNSS receiver's pseudorange, carrier phase and carrier Doppler shift observations are commonly modeled as

$$\rho_i = r_i + c(\Delta t_{rx} + \Delta t_{SV,i}) + \Delta r_{ion,i} + \Delta r_{trop,i} + \Delta r_{eph,i} + \Delta r_{MP,\rho,i} + \Delta r_{HW,\rho,i} \tag{1}$$

$$\Phi_i = r_i + c(\Delta t_{rx} - \Delta t_{SV,i}) - \Delta r_{ion,i} + \Delta r_{trop,i} + \Delta r_{eph,i} + \lambda_i N_i + \Delta r_{MP,\Phi,i} + \Delta r_{HW,\Phi,i}$$
(2)

$$v_{Do,i} = \dot{r}_{Rx,i} + c(\Delta \dot{t}_{rx} - \Delta \dot{t}_{SV,i}) - \Delta \dot{r}_{ion,i} + \Delta \dot{r}_{trop,i} + \Delta \dot{r}_{MP,\Phi,i} + \Delta \dot{r}_{eph,i} + \Delta v_{HW,Do,i}$$
(3)

with

$$\dot{r}_{Rx,i} = \dot{r}_i - \dot{r}_{SV,i}$$

where  $\rho_i$  is the pseudorange observation,  $r_i$  the slant range according to the broadcast ephemeris, c the nominal speed of light used by the GNSS system,  $\Delta t_{rx}$  the receiver clock offset,  $\Delta t_{SV,i}$  the satellite clock offset,  $\Delta r_{ion,i}$  the range error due to ionospheric diffraction,  $\Delta r_{trop,i}$  the range error due to tropospheric diffraction,  $\Delta r_{eph,i}$  the line of sight (LOS) broadcast ephemeris error,  $\Delta r_{MP,\rho,i}$ the range error due to code multipath,  $\Phi_i$  the carrier phase range observation,  $\lambda$  the nominal carrier phase wavelength, N the carrier cycle ambiguity,  $\Delta r_{MP,\Phi,i}$  carrier phase multipath,  $v_{Do,i}$  the Doppler shift observation, transformed to a velocity and  $\dot{r}_{Rx,i}$  the slant range rate corrected for the satellite's LOS velocity. Hardware noise is represented by  $\Delta r_{HW,\rho,i}$ ,  $\Delta r_{HW,\Phi,i}$  and  $\Delta v_{HW,Do,i}$  respectively.  $\dot{r}_{SV,i}$ is the range rate due to satellite motion as computed from broadcast ephemeris.

Differential augmentation systems exploit the fact that a major part of the measurement errors is highly correlated between two receivers that are sufficiently close to one another, due to the fact that the signals they receive experience very similar conditions on their way through the atmosphere. This allows the estimation of the common errors once the position of one receiver - herein referred to as the base station - is known.

## **3** Base station

The base station consists of a GPS receiver that is placed at a known location and a data processing system that computes the corrections and broadcasts them via a wireless datalink to an unlimited amount of UAS.

Both code and phase observations evolve rapidly in time due to the high orbital velocity of the satellites (affecting  $r_i$ ) and the rapidly drifting clock offset  $\Delta t_{rx}$  of the u-blox receiver's free running clock. While the latter is common to all observations and thus is eliminated by solving the navigation equation, the first imposes hard timing constraints on the data link and the synchronicity of the measurements of the base receiver and the airborne receiver when using a single-differencing approach. Therefore,  $r_i$  is computed from broadcast ephemerides (see [2]) and subtracted from the pseudorange observations to form estimates of the common range errors. The base receiver's clock error is not removed from the corrections to avoid increasing the variance of the corrections. Thus the **pseudorange corrections** are formed as

$$\rho_{c,i} = c(\Delta t_{rx} + \Delta t_{SV,i}) + \Delta r_{ion,i} + \Delta r_{trop,i} + \Delta r_{eph,i} + \Delta r_{MP,\rho,i} + \Delta r_{HW,\rho,i}$$
(4)

The phase observations are treated in the same way, forming **carrier phase corrections**:

$$\Phi_{c,i} = c(\Delta t_{rx} - \Delta t_{SV,i}) - \Delta r_{ion,i} + \Delta r_{trop,i} + \Delta r_{eph,i} + \lambda_i N_i + \Delta r_{MP,\Phi,i} + \Delta r_{HW,\Phi,i}$$
(5)

3

Doppler corrections are obtained by time-differencing of carrier phase corrections. Note that due to the fact that the broadcast ephemerides are only local approximations of the satellite orbits, it has to be made sure that the base station and the airborne receiver use the same set of ephemerides.

### 3.1 Base station positioning

It will seldom be the case that a surveyed reference point is available for the base station. Consequently, a way has to be found to estimate its position in a reasonable amount of time. An error of the base position directly translates into an additional position offset of receivers using its corrections, thus the quality of the ground station positioning determines the achievable accuracy for the UAS. The International GNSS Service (IGS) provides observation data of a worldwide network of GNSS reference stations. A subset of these stations provide their data in close-to-real-time via NTRIP streams. Tests have been run and it could be demonstrated that, using the open-source RTK positioning software rtklib, the base receiver position can be computed with sufficient accuracy within 20 minutes, provided there is a reference station that is sufficiently close. Please see section 5.1 for experimental time-to-ambiguity-resolution results for three baselines up to 134 km.

## 4 Airborne positioning algorithm

The airborne component applies the received corrections to its own observations, canceling the major part of common errors, leading to the corrected observations

$$\rho_{i,c} = c\Delta t_{rx,c} + \Delta r_{MP,\rho,i} + \Delta r_{HW,\rho,i,c} + \epsilon \Delta r_{\rho,c} \tag{6}$$

$$\Phi_{i,c} = c\Delta t_{rx,c} + \lambda N_c + \Delta r_{MP,\Phi,i} + \Delta r_{HW,\Phi,i,c} + \epsilon \Delta r_{\Phi,c}$$
(7)

$$v_{Do,i,c} = c\Delta t_{rx} + \epsilon \Delta r_{\Phi,c} + \Delta v_{HW,Do,i,c} + \epsilon \Delta r_{Do,c} \tag{8}$$

where  $\Delta t_{rx,c}$  is the difference of the clock offsets of base and airborne receiver,  $N_c$  the difference of their integer ambiguities,  $\Delta r_{HW,\rho,i,c}$ ,  $\Delta r_{HW,\Phi,i,c}$  and  $\Delta v_{HW,Do,i,c}$  the combined noises of observations and corrections. Residual common errors due to timely and spatial de-correlation are represented by  $\epsilon \Delta r_{\rho,c}$ ,  $\epsilon \Delta r_{\Phi,c}$  and  $\epsilon \Delta r_{Do,c}$ .

The UAS position is obtained in two steps. First, a relative trajectory based on carrier phase time differences and integrated Doppler observations is computed. Second, the absolute position is obtained from pseudoranges and both are merged by an indirect Kalman Filter. Both steps are described in more detail in the following sections.

Position domain code smoothing is superior to range domain approaches under dynamic conditions for two major reasons: First, three coordinates to be estimated instead of up to e.g. about seven ranges for GPS leads to smaller filter matrices and thus reduces execution time.

Second, and more importantly, in the range domain, after loss of lock and reacquisition of a satellite, its smoothed range is lost and the respective filter channel is re-initialized. Afterwards, the smoothed range will be unusable while the filter re-converges. This can be impractical if less than 4 converged ranges are available, forcing the rover to use a range being still in the convergence phase and leading to a degraded position estimate.

On the other hand, in the position domain only an arbitrary set of four or more range and delta range observations has to be available at all times to keep up a continuous smoothed position. A reacquired satellite can be used immediately for positioning. This is a particularly valuable property in urban environments. Imagine an airborne rover that subsequently looses lock to all satellites for a few epochs, but manages to maintain a set of 4 satellites all the time. If the rover uses a range domain filter, it is left with a set of filtered ranges whose error is thrown back to that of their pseudorange measurements. A position domain filter might experience an increased noise level due to a higher PDOP associated with fewer satellites in view, but run continuously.

#### 4.1 Background Trajectory

Numerous approaches exist to solve for the unknown integer ambiguities to obtain a highly accurate purely carrier phase based position. Most of them are search-based (e.g. LAMBDA, introduced by [8]), but there are promising analytical approaches for On-The-fly ambiguity resolution as well, e.g. [3]. There is no known technique that provides reliable and instantaneous ambiguity resolution for L1-only receivers though. The author of [5] shows, based on theoretical considerations, that at least 12 satellites in view are necessary for on-the-fly ambiguity resolution in 100% of all cases, a condition that is not met by the GPS constellation.

A simple alternative to exploit the excellent accuracy of carrier phase measurements without the need for ambiguity resolution consists in time-differencing. Since for a continuously tracked signal the ambiguities are constant in time, they are canceled by subtracting the observations of two subsequent epochs. The method has been successfully applied to standalone L1 phase observations for the analysis of flight maneuvers of aircrafts and birds (see [10], [11]).

As every technique involving phase observations, this method is susceptible to cycle slips and loss of the carrier signal. In situations where - e.g. due to steep maneuvers - the number of available phase observations drops below four, the accurately tracked trajectory is lost.

To overcome this drawback, the method has been extended by incorporating integrated Doppler observations that are less susceptible to dynamic stress, leading to better availability. Whenever the number of available carrier phase time differences drops below the minimum of four, they are augmented by integrated Doppler observations, i.e.

 $\delta \mathbf{r} = \begin{bmatrix} \delta r_{cp,1} \\ \vdots \\ \delta r_{cp,n_{cp}} \end{bmatrix} \quad \text{for } n_{cp} \ge 4 \tag{9}$   $\delta \mathbf{r} = \begin{bmatrix} \delta r_{cp,1} \\ \vdots \\ \delta r_{cp,n_{cp}} \\ \delta r_{Do,1} \end{bmatrix} \quad \text{for } n_{cp} < 4 \tag{10}$ 

$$\begin{bmatrix} \vdots \\ \delta r_{Do,(n_{cp}-4)} \end{bmatrix}$$

with

$$\delta r_{Do,i} = \int_{t_{k-1}}^{t_k} v_{Do,i} \,\mathrm{d}t \tag{11}$$

Where the vector  $\delta \mathbf{r}$  holds the delta ranges used to compute the displacement between two epochs,

 $\delta r_{cp,i}$  are carrier phase time differences,  $\delta r_{Do,i}$  are integrated Doppler delta ranges of satellites whose carrier phase time differences are not available and  $n_{cp}$  represents the number of available carrier phase time differences.

Doppler observations are a snapshot of the derivative of the trajectory, i.e. the velocity, at the time of measurement. Their integral therefore only approximates the true displacement and integrated Doppler ranges are noisier than carrier phase ranges. The advantage of not loosing track of the trajectory however largely outweighs this drawback, the more so as integrated Doppler observations are only used during the short periods of carrier phase outages.

The position increment between two epochs is calculated by linearizing the observation equation (here given for one range, note that  $\delta r_{i,k}$  represents the carrier phase time differences augmented by integrated Doppler delta ranges)

$$\delta r_{i,k} = \sqrt{(x_k^{SV} - x_k^{Rx})^2 + (y_k^{SV} - y_k^{Rx})^2 + (z_k^{SV} - z_k^{Rx})^2} - \sqrt{(x_{k-1}^{SV} - x_{k-1}^{Rx})^2 + (y_{k-1}^{SV} - y_{k-1}^{Rx})^2 + (z_{k-1}^{SV} - z_{k-1}^{Rx})^2} + c(\Delta t_{Rx,k} - \Delta t_{Rx,k-1}) + (\epsilon_k - \epsilon_{k-1})$$
(12)

and by solving it iteratively for  $x_k$  by the method of least squares.

#### 4.2 Absolute positioning

Based on the corrected pseudoranges, the absolute position is computed following the standard procedure of linearizing and iteratively solving the navigation equation. To correct for the signal propagation delay, pseudoranges are time-tagged at the receiver, as proposed by [1]. The resulting noisy position estimate is used to correct the position offset of the background trajectory by means of a simple linear Kalman filter.

The prediction-correction structure of a Kalman Filter usually requires a dynamic model of the system whose state is to be estimated and knowledge about the control inputs. In this case, the offset of the background trajectory from the true trajectory drifts very little over time thanks to its excellent relative precision. It can therefore be approximately modeled as being constant, i.e the dynamics matrix  $\mathbf{A} \approx I_3$  which is equivalent to omitting the state prediction step.

This naturally leads to an indirect i.e. error state filter formulation, as the difference between the pseudorange trajectory and the background trajectory is a noisy measurement of the background trajectory offset. The filter equations become

$$\Delta \boldsymbol{x}_k = \boldsymbol{x}_{\boldsymbol{b},\boldsymbol{k}} - \Delta \boldsymbol{x}_{\boldsymbol{true},\boldsymbol{k}} \tag{13}$$

$$\Delta \hat{\boldsymbol{x}}_{k}^{-} = \boldsymbol{A} \boldsymbol{\Delta} \hat{\boldsymbol{x}}_{k-1} \tag{14}$$

$$\boldsymbol{P}_{k}^{-} = \boldsymbol{A}\boldsymbol{P}_{k-1}\boldsymbol{A}^{T} + \boldsymbol{Q} \tag{15}$$

$$\Delta \boldsymbol{z}_{k} = \boldsymbol{x}_{\rho,k} - \boldsymbol{x}_{b,k} \tag{16}$$

$$\mathbf{K}_{k} = \mathbf{F}_{k} \left( \mathbf{F}_{k} + \mathbf{R} \right) \tag{11}$$

$$\Delta \hat{\mathbf{x}}_{k} = \Delta \hat{\mathbf{x}}^{-} + \mathbf{K}_{k} \left( \Delta \hat{\mathbf{x}}_{k} - \Delta \mathbf{x}_{k} \right) \tag{18}$$

$$\Delta x_k = \Delta x_k + \mathbf{K}_k (\Delta x_k - \Delta z_k) \tag{18}$$

$$\boldsymbol{P}_k = (\boldsymbol{I} - \boldsymbol{K}_k) \boldsymbol{P}_k \tag{19}$$

 $\Delta \boldsymbol{x}_k$  being the error state,  $\Delta \boldsymbol{z}_k$  being the error state measurement,  $\boldsymbol{K}_k$  being the filter gain,  $\Delta \hat{\boldsymbol{x}}_k^-$ ,  $\Delta \hat{\boldsymbol{x}}_k$  being the *a priori* and *a posteriori* error state estimate respectively,  $\boldsymbol{P}_k^-$ ,  $\boldsymbol{P}_k$  being the *a priori* and *a posteriori* estimation covariance matrix respectively,  $\boldsymbol{Q}$  being the process noise and  $\boldsymbol{R}$  being the measurement noise.

## 5 Experimental results

The system has been implemented in Matlab/Simulink and applied to observations collected on a rotating test bench on the ground. Flight tests have been performed as well but are not considered in this article since no RTK trajectory could be obtained in post processing. Base positioning using freely available close-to-real-time observations by the IGS network ([6]) has been tested for three baseline lengths up to 134 km.

**Hardware** Data have been collected with GPS receivers of type u-blox LEA-6T. The LEA-6T chipset is one of the least expensive on the market providing raw observations on the GPS L1 band. Receiver modules integrating the same receiver chip as the evaluation kit used for our experiments are available with a weight of 16g including an active antenna, making them suitable for small UAS down to a few hundred grams TOW. The rover and base antennas where an active patch antenna (u-blox ANN-MS-0-005) and a survey grade antenna on a surveyed reference position, respectively.

#### 5.1 Base positioning

To verify the assumption that the position of the base receiver can be determined in a reasonable amount of time using freely available reference observations and open source software, tests for three successively large baselines have been performed. Data have been collected for 20 minutes and processed with rtklib. The positions obtained by averaging all positions with fixed ambiguities are of centimeter-level quality (see Table 1), better accuracy can be expected for longer sampling periods. Longer baselines have been tested but did not lead to fixed ambiguities within 20 minutes.

$b \; [\mathrm{km}]$	$t_{fix}$ [s]	$n_{fix}[\%]$	$\sigma_N \; [\rm{mm}]$	$\sigma_E \; [\rm{mm}]$	$\sigma_D \; [\rm{mm}]$	$\mu_N \text{ [mm]}$	$\mu_E \text{ [mm]}$	$\mu_D \; [\rm{mm}]$
8	403	68	4.2	3.7	9.4	2.2	-1.4	-19.9
64	96	9	1.8	3.6	10.6	0.8	1.3	1.7
134	187	53	12.2	11.5	22.6	-4.4	13.2	8.7

Table 1: LEA-6T base positioning with rtklib for different baselines (stations TLMF, PAYR, NARB).  $t_{fix}$  is the time when ambiguities where fixed for the first time,  $n_{fix}$  the percentage of epochs for which ambiguities could be fixed within 20 minutes.

### 5.2 Dynamic tests on the ground

To emulate flight conditions, the receiver has been placed on a freely rotating turntable with a radius of 4.2 m. As with all localization systems, a key challenge consists of the comparison of measurements from the system to be tested versus a known truth source. In this case the true trajectory could be reconstructed with RTK-level accuracy using the open-source software rtklib (see e.g. [7]). The bar

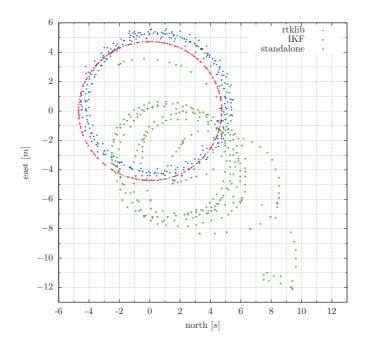


Figure 1: North and east coordinate of trajectory in local NED frame

was rotated by one person. A second receiver was set up to record raw data at the campus' surveyed reference antenna. In parallel to the raw observations, the standalone position solution of the receiver has been logged. The test site was located 40 to 80 m from two two-story buildings, thus the rover receiver was likely to experience multipath.

The position errors over time are given in Fig. 2, see Fig. 3 and Tab. 2 for error statistics.

While not exceeding 5 m most of the time, the standalone error rises to up to 16 m in the east coordinate around 330 s. The filtered differential solution error stays well close to 1 m most of the time and exhibits a smaller variance. Note also that the filtered solution shows no sign of the two discontinuities at 25 s and 305 s in the receiver's solution. Regarding the total error (norm of the error vector), for the standalone solution 95 % of all 3D errors are smaller than 11.15 m. The Kalman filtered differential solution brings the 95% error range down to 1.35 m.

	$\sigma_N$ [m]	$\sigma_E$ [m]	$\sigma_D$ [m]	$\mu_N  [\mathrm{m}]$	$\mu_E  [\mathrm{m}]$	$\mu_D$ [m]	HDOP	VDOP	
Circular trajectory, baseline 180 m									
standalone	2.18	2.06	1.74	2.67	-4.41	0.74	-	-	
Differential C/A	0.86	0.65	2.39	0.06	-0.2	-0.47	1.2 - 1.4	2.1 - 3.6	
IKF	0.21	0.21	0.26	0.38	0.49	-0.85	1.2 - 1.4	2.1 - 3.6	

Table 2: Position error standard deviations  $\sigma$  and means  $\mu$  under static and dynamic conditions (300 s, 1 Hz)

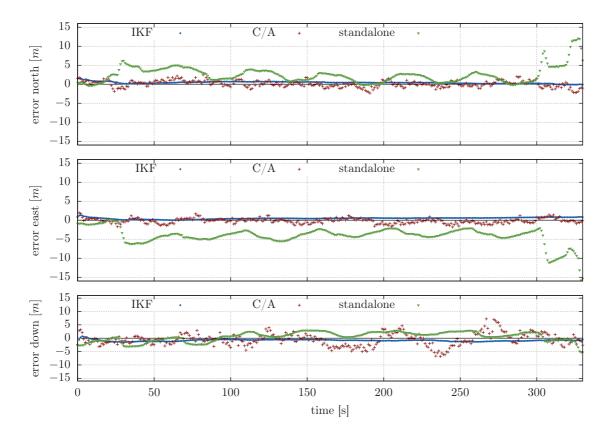


Figure 2: Position offset to RTK trajectory in local NED system: indirect Kalman Filter, differential C/A, LEA-6T standalone solution (SBAS enabled, configured for airborne platforms  $\leq 4$  g)

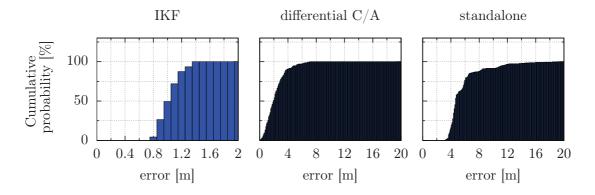


Figure 3: Cumulative distributions of 3D error on circular trajectory. Note the different scales of the x axes.

## 6 Conclusions and future work

This work has investigated the feasibility of a local GPS augmentation system based on low cost hardware components. A code smoothing filter has been developed that takes advantage of the complementary qualities of carrier phase, Doppler and pseudorange observations, based on an indirect linear Kalman filter. The positioning filter has been implemented in Matlab/Simulink and applied to observation data collected on a rotating test bench. A highly accurate reference trajectory could be reconstructed using open - source RTK post-processing software. The results are quite promising and reveal the shortcomings of standalone dynamic GPS positioning. Differential positioning shows an improvement of the mean position error (the only exception being the vertical error whose mean values are of similar magnitude) as well as the error variance by roughly one magnitude. Further efforts will be focused on the implementation of the proposed system on an embedded system and the verification of its performance under flight conditions, including the tracking of the reference trajectory by means of GPS-independent instruments such as theodolites to exclude the possibility of common errors. Its robustness thanks to the integration of Doppler measurements makes the proposed system an excellent basis and safety fall-back for the evaluation of more accurate but less reliable techniques involving ambiguity resolution.

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