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1	Open Source MATLAB Code for GPS Vector Tracking on a Software-Defined
2	Receiver
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6	
7	ABSTRACT
8	The research regarding Global Positioning System (GPS) vector tracking (VT), based
9	on a software-defined receiver (SDR), has been increasing in recent years. The
10	strengths of VT include its immunity to signal interference, its capability to mitigate
11	multipath effects in urban areas, and its excellent performance in tracking signals under
12	high-dynamic applications. We developed open source MATLAB code for GPS VT
13	SDR to enable researchers and scientists to investigate its pros and cons in various
14	applications and under various environments. To achieve this goal, we developed an
15	"equivalent conventional tracking (CT)" SDR as a baseline to compare with VT. The
16	GPS positioning estimator of this equivalent CT is based on an extended Kalman filter
17	(EKF), which has exactly the same state, system and carrier measurement models and
18	noise tuning method as VT. This baseline provides users with a tool to compare the
19	performance of VT and CT on common ground. In addition, this MATLAB code is
20	well-organized and easy to use. Users can quickly implement and evaluate their own
21	newly developed baseband signal processing algorithms related to VT. The
22	implementation of this VT code is described in detail. Finally, static and kinematic
23	experiments were conducted in an urban and open-sky area, respectively, to show the
24	usage and performance of the developed open source GPS VT SDR.
25	

Keywords GPS, Software-defined receiver (SDR), Vector tracking (VT), Open-source
 software, Extended Kalman filter (EKF)

28

## 29 Introduction

30 Reliable navigation is highly desirable in challenging environments where navigation satellite signals are interfered with and attenuated. To obtain a navigation solution, 31 32 satellite signals must be tracked continually so that the ephemeris data can be decoded 33 and the measurements (such as pseudoranges and pseudorange rates) can be extracted. In conventional Global Positioning System (GPS) receivers, each acquired satellite is 34 35 allocated to an individual tracking channel. Each channel has two closed loops, one for code and one for carrier. All tracking channels are independent of each other, i.e., there 36 37 is no interaction between channels, and no information exchange between signal 38 tracking and navigation processors. In vector tracking (VT)-based receivers, tracking channels are coupled together through the navigation processor, often based on an 39 extended Kalman filter (EKF). Different forms of Kalman filter implementation can be 40 41 found in (Won et al. 2010). The fundamental principle behind VT is the relationship between the code or carrier phase and the receiver states of position, velocity and time 42 (PVT), which was first proposed by Copps in the early 1980s (Copps et al. 1980). The 43 vector delay lock loop (VDLL) is described in (Spilker 1996), where the code is tracked 44 in the vector mode, while the carrier tracking remains the same as in the conventional 45 receiver. The boom of computer technologies and inertial devices has pushed the 46 47 development and application of vector tracking in the last two decades.

Previous research has mainly focused on the advantages of vector tracking over 48 conventional tracking. The most commonly cited benefits are its increased capabilities 49 in harsh environments, e.g., low carrier-to-noise ratio (CNR) (Lashley and Bevly 2009; 50 51 Lashley et al. 2009; Pany and Eissfeller 2006), intermittent signal outages (Lashley and Bevly 2007; Zhao and Akos 2011; Zhao et al. 2011), and high dynamics (Lashley et al. 52 2009), due to the mutual aiding of the channels with respect to each other and a higher 53 filtering gain to be used stably (Groves and Mather 2010). To further improve 54 55 robustness and accuracy in poor environments, vector tracking can be easily integrated 56 with an inertial navigation system (INS) by simply augmenting the navigation Kalman filter with appropriate INS-related states (Lashley and Bevly 2013; Luo et al. 2012; 57

Petovello and Lachapelle 2006). In recent years, with the increasing development of 58 59 intelligent transportation systems and location-based services in urban canyon areas, 60 vector tracking has received more attention. For example, in (Hsu et al. 2013; Hsu et al. 2015b; Syed Dardin et al. 2013), vector tracking is applied to multipath or non-line-of-61 sight reception mitigation in the signal processing stage, while in (Ng and Gao 2017) 62 63 deeply coupled multi-receiver vector tracking is used to improve the reliability and robustness of GPS signal tracking and position estimation. A more recent paper 64 converts a software defined receiver (SDR) to a signal simulator by using vector 65 tracking loop to create desired line-of-sight parameters for updating the numerically 66 controlled oscillator (NCO) and therefore generate the code and carrier replicas (Maier 67 et al. 2018). Apart from the benefits and applications mentioned above, vector tracking 68 has also been used to improve bit synchronization and decoding (Ren et al. 2013), 69 estimate ionosphere residual error (Shytermeja et al. 2017), enhance carrier phase 70 tracking (Brewer and Raquet 2016), etc. The idea of vector tracking also yields other 71 signal tracking techniques, e.g., direct position tracking loops (Liu et al. 2011) and 72 73 robust adaptive joint tracking (Tabatabaei and Mosavi 2017). It should be noted that the coupling of loops is not only responsible for vector tracking's superior performance, 74 but also allows error propagation among loops, which has been dealt with in 75 (Bhattacharyya and Gebre-Egziabher 2010; Sun et al. 2016). 76

77 The majority of the current research generally focuses on the exploration of 78 benefits offered by vector tracking, but seldom presents the detailed implementation of 79 vector tracking. In 2011, Zhao and Akos (Zhao and Akos 2011) published an open source code of vector tracking based on the GPS software defined receiver developed 80 81 by Borre et al. (Borre et al. 2007), which is a popular open source SDR platform for 82 beginners. In Zhao's open-source software, the performance of vector tracking is 83 compared with that of traditional scalar loops and navigation solutions estimated using the least squares method. In fact, improvements of vector tracking might be due to the 84 Kalman filter; an equivalent conventional receiver must be implemented as a reference. 85 86 In this paper, a fully self-developed SDR based on vector tracking is presented. An

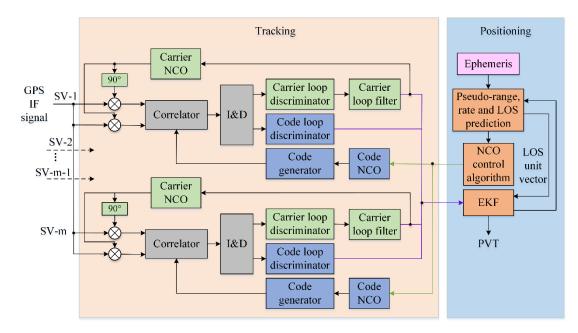
equivalent conventional tracking (CT)-based SDR using delay lock loop (DLL) and
phase lock loop (PLL) is also implemented for performance comparison between VT
and CT. The CT-based SDR uses an EKF to estimate receiver's PVT. The system
propagation and measurement model and the noise tuning method are exactly the same
for VT and CT. This feature can bring them both to common ground to allow an accurate
performance evaluation.

In the following sections, the design of vector tracking in the open source SDR is described first. Afterward, main functionalities of the software are given. Then, the experiments are conducted to evaluate the performance of this software. Finally, conclusions are drawn, including future work.

97

## 98 Vector tracking algorithm

99 In this SDR, VDLL is implemented as an example. Users can easily extend this software 100 to vector frequency lock loop (VFLL), or vector delay/frequency lock loop (VDFLL). 101 Fig. 1 presents the architecture of this SDR. As shown in Fig. 1, each acquired satellite in the incoming intermediate frequency (IF) signal is allocated to one tracking channel. 102 103 In each channel, IF signals are first multiplied with the locally generated carrier replica 104 in both in-phase and quadrature arms. Correlation is then performed between the code replicas and the received ones. In this paper, three code replicas spacing of 0.5 chips 105 are generated. Afterwards, correlation results are integrated and dumped. The output of 106 107 these integrations is used as the input to the carrier/code loop discriminator to find the phase error of the local carrier and code replicas. In each carrier loop, the carrier 108 109 discriminator output is filtered and fed back to the carrier NCO, so as to modify the 110 frequency of local carrier replica. For the code tracking loop, code discriminator outputs 111 of all channels are forwarded to the navigation processor. In this paper, an EKF is used. 112 The output of the carrier loop filter, i.e., Doppler shift frequency information, is also 113 fed into the EKF. Note that in practice the EKF update time is not necessary to be the same as the coherent integration time (typically 1 ms for GPS L1 signal). A pre-filter 114 can be used to average the code discriminator outputs over multiple integration time, 115







**Fig. 1** The tracking architecture of the developed GPS VT SDR

119 The EKF estimates the receiver PVT based on its system propagation and the measurements, which will be described in detail later. After obtaining the navigation 120 121 solution, the pseudorange and its rate and the line-of-sight (LOS) vector between the 122 receiver and the satellites are predicted. To do this, the satellite ephemeris data must be 123 known a priori. In this paper, conventional tracking is used to process the IF signal and 124 decode the ephemeris data first. The PVT calculated using conventional tracking is then 125 used to initialize the VDLL. Finally, the predicted pseudo-ranges are used to control the 126 code NCO and then are fed back to each channel.

127

## 128 Design of the Extended Kalman Filter

129 The state vector of the EKF is:

130 
$$\mathbf{X} = \begin{bmatrix} \Delta p_x, \Delta p_y, \Delta p_z, \Delta v_x, \Delta v_y, \Delta v_z, \Delta b, \Delta d \end{bmatrix}^T$$
(1)

131 where  $\Delta \mathbf{p} = [\Delta p_x, \Delta p_y, \Delta p_z]$  and  $\Delta \mathbf{v} = [\Delta v_x, \Delta v_y, \Delta v_z]$  are the three-dimensional 132 receiver position and velocity error vectors in an earth-centered and earth-fixed (ECEF) 133 frame;  $\Delta b$  and  $\Delta d$  are the receiver clock bias and drift errors in the units of meters and 134 meters per second, respectively. The system propagation at epoch k is:

135 
$$\hat{\mathbf{X}}_{k}^{-} = \boldsymbol{\Phi}_{k-1} \hat{\mathbf{X}}_{k-1}^{+}$$
(2)

136 where

137 
$$\Phi_{k-1} = \begin{bmatrix} \mathbf{I}_{3\times3} & \tau \mathbf{I}_{3\times3} & \mathbf{0}_{3\times2} \\ \mathbf{0}_{3\times3} & \mathbf{I}_{3\times3} & \mathbf{0}_{3\times2} \\ \mathbf{0}_{2\times3} & \mathbf{0}_{2\times3} & \mathbf{K} \end{bmatrix}_{8\times8}$$
(3)

138 
$$\mathbf{K} = \begin{bmatrix} 1 & \tau \\ 0 & 1 \end{bmatrix}$$
(4)

In equation (3), *τ* is the update interval of the EKF. The superscript and subscript,
"-" and "+", denote the system state before and after measurement update, respectively.
The symbol "^" represents the EKF estimates.

142 The measurements of the EKF are the pseudo-range error,  $\Delta \rho^{j}$ , and pseudo-range 143 rate error,  $\Delta \dot{\rho}^{j}$ , of satellite *j*. The pseudo-range error is

144 
$$\Delta \rho^{j} = \Delta \tau^{j} \cdot \frac{c}{f_{CA}}$$
(5)

145 where  $\Delta \tau^{j}$  is the code discriminator output in chips,  $f_{CA}$  is the code chipping rate 146 (1.023 MHz for GPS L1 C/A); *c* is the speed of light. The error of pseudo-range rate 147 is the difference between the measured pseudo-range rates extracted from the carrier 148 tracking loop and the predicted ones calculated using the estimated receiver velocity 149 and satellite velocity as well as the estimated receiver clock drift.

150 
$$\Delta \dot{\rho}^{j} = f_{Doppler}^{j} \cdot \frac{c}{f_{L1}} - \left(\mathbf{v}_{usr} - \mathbf{v}_{sate}^{j}\right) \cdot \mathbf{l}^{j} - \hat{d}_{u,clk} + d_{sv,clk}^{j}$$
(6)

where  $f_{Doppler}^{j}$  is the Doppler shift frequency in Hz;  $f_{L1}$  is the carrier frequency (1575.42 MHz for GPS L1);  $\mathbf{v}_{usr}$  and  $\mathbf{v}_{sate}^{j}$  are the velocity vectors of the receiver and satellite j, respectively;  $\mathbf{l}^{j}$  is the LOS unit vector from the receiver to satellite j;  $\hat{d}_{u,clk}$  and  $d_{sv,clk}^{j}$  are the estimated receiver clock drift and the  $j^{\text{th}}$  satellite clock drift, respectively, both in meters per second. The measurement vector can be expressed as

156 
$$\mathbf{Z} = \left[\Delta \rho^{j}, \Delta \dot{\rho}^{j}\right]$$
(7)

157 The relationship between the state vector and the measurement vector at epoch k is 158 linearized by a first-order Taylor's expression as follows

159 
$$\mathbf{Z}_{k} = \mathbf{H}_{k} \cdot \mathbf{X}_{k}$$
(8)

160 where **H** is the measurement matrix, calculated as

161 
$$\mathbf{H} = \begin{bmatrix} -\mathbf{l}_{x}^{1} & -\mathbf{l}_{y}^{1} & -\mathbf{l}_{z}^{1} & 0 & 0 & 0 & 1 & 0 \\ -\mathbf{l}_{x}^{2} & -\mathbf{l}_{y}^{2} & -\mathbf{l}_{z}^{2} & 0 & 0 & 0 & 1 & 0 \\ \vdots & \vdots \\ -\mathbf{l}_{x}^{m} & -\mathbf{l}_{y}^{m} & -\mathbf{l}_{z}^{m} & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & -\mathbf{l}_{x}^{1} & -\mathbf{l}_{y}^{1} & -\mathbf{l}_{z}^{1} & 0 & 1 \\ 0 & 0 & 0 & -\mathbf{l}_{x}^{2} & -\mathbf{l}_{z}^{2} & 0 & 1 \\ \vdots & \vdots \\ 0 & 0 & 0 & -\mathbf{l}_{x}^{m} & -\mathbf{l}_{y}^{m} & -\mathbf{l}_{z}^{m} & 0 & 1 \end{bmatrix}.$$
(9)

where *m* is the number of satellites involving positioning; the subscript of the LOS unit vector denotes its x, y, and *z* components, and the superscript denotes the satellite.

164

## 165 Noise Tuning of the EKF

The process noise comes from two sources, i.e., the receiver dynamics and clock noise,as follows

168 
$$\mathbf{Q} = \begin{bmatrix} \mathbf{Q}_{dyn} & \mathbf{0}_{6\times 2} \\ \mathbf{0}_{2\times 6} & \mathbf{Q}_{clk} \end{bmatrix}$$
(10)

169 The values of  $\mathbf{Q}_{dyn}$  and  $\mathbf{Q}_{clk}$  can be set empirically according to the receiver motion 170 state and the oscillator used. Alternatively, they can be calculated as

171 
$$\mathbf{Q}_{dyn} = \begin{bmatrix} \tau^3 / 3 \cdot \mathbf{I}_{3\times 3} & \tau^2 / 2 \cdot \mathbf{I}_{3\times 3} \\ \tau^2 / 2 \cdot \mathbf{I}_{3\times 3} & \tau \cdot \mathbf{I}_{3\times 3} \end{bmatrix} \cdot S_{\nu}$$
(11)

172 
$$\mathbf{Q}_{clk} = \begin{bmatrix} S_f \cdot \tau + S_g \tau^3 / 3 & S_g \tau^2 / 2 \\ S_g \tau^2 / 2 & S_g \cdot \tau \end{bmatrix}$$
(12)

173 where  $S_v$  is the receiver velocity noise power spectral density (PSD);  $S_f$  and  $S_g$  are the 174 PSD of receiver clock phase and frequency, respectively. The value of  $S_v$  should be set 175 according to the level of dynamics. Settings of  $S_f$  and  $S_g$  are usually based on the rule 176 of thumb values of the type of oscillator used, or calculated using the following 177 formulas

$$S_f = c^2 \cdot \frac{h_0}{2} \tag{13}$$

179 
$$S_g = c^2 \cdot 2\pi^2 \cdot h_{-2}$$
 (14)

180 where  $h_0$  and  $h_{-2}$  are the coefficients of white frequency modulation noise and flicker 181 frequency modulation noise of the oscillator used, respectively.

182 The measurement noise covariance matrix is calculated adaptively using the 183 innovation-based adaptive estimation technique (Mohamed and Schwarz 1999). The 184 measurement innovation at epoch k + 1 in this paper is

185 
$$\mathbf{V}_{k+1} = \mathbf{Z}_{k+1} - \mathbf{Z}_{k+1}^-$$
 (15)

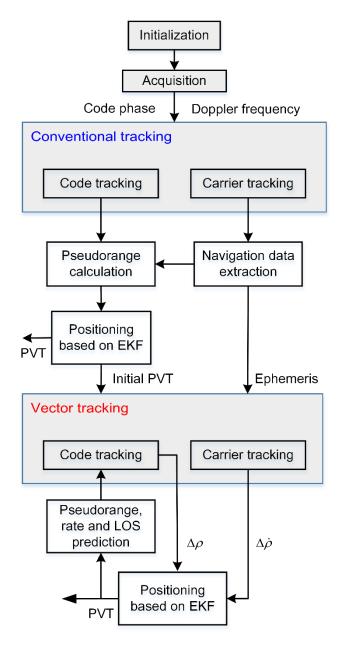
186 
$$\mathbf{Z}_{k}^{-} = \mathbf{H}_{k} \hat{\mathbf{X}}_{k}^{-}$$
(16)

187 The diagonal element of the measurement covariance matrix is the variance of the 188 measurement innovation. The off-diagonal terms are assumed to be zero due to the 189 weak correlation between channels.

190

## 191 Main Functionalities of the Open-Source SDR

This open-source SDR is developed using MATLAB, which is an easy-to-use programming language, so that users can focus more on the implementation of the newly developed algorithms. Fig. 2 presents the flowchart of the software. The four main functionalities include initialization, acquisition, conventional tracking and vector tracking, which are described in detail as follows:



197

198

Fig. 2 Flowchart of the open-source GPS SDR

## 199 Initialization

200 The first step to use this software is to complete configurations such as the sampling

201 rate and intermediate frequency of the raw signal, the frequency step and band to be

searched in the acquisition, etc.

#### 203 Acquisition

The second module is signal acquisition, which determines code phase and Doppler frequency of visible satellites. A two-step coarse-to-fine acquisition method is used. In the first step, 4-ms data is used to detect the code phase and Doppler frequency coarsely via the parallel code phase search acquisition algorithm (Van Nee and Coenen 1991). The second step utilizes long C/A code-stripped data to find the carrier frequency accurately via the fast Fourier transformation technique.

210

226

## 211 Conventional Tracking

212 After obtaining the code phase and Doppler frequency, these two parameters should be 213 refined in the tracking stage so that satellite ephemeris data can be decoded. Measurements of pseudorange and pseudorange rate can also be obtained during 214 215 tracking. A second-order DLL and PLL is used in this software. With this information, 216 the navigation solution is calculated in the positioning module, which is based on an 217 EKF instead of the least-squares method in this SDR, because any improvements of vector tracking might be due to the Kalman filter. The EKF used in the conventional 218 219 receiver has the same states, system and measurement models as the vector tracking 220 EKF. The noise tuning of these two EKFs are also the same so as to compare the 221 performance of the conventional and vector tracking methods based on common ground. 222 Even so, there still exist two differences between the conventional tracking and vector 223 tracking. One difference is the formation of pseudorange error measurements. In conventional tracking, it is calculated by the measured pseudorange minus the predicted 224 225 pseudorange as follows

$$\Delta \rho^{j} = c \left( t_{rx} - t_{tx}^{j} \right) - \left\| \mathbf{r}_{u} - \mathbf{r}^{j} \right\| - \hat{b}_{clk}$$
(17)

where  $t_{rx}$  is the receiver time in a conventional receiver;  $t_{tx}^{j}$  is the transmission time from satellite *j*;  $\mathbf{r}_{u}$  and  $\mathbf{r}^{j}$  are the position of receiver and satellite *j*, respectively;  $\hat{b}_{clk}$  is the estimated receiver clock bias. In vector tracing, however, the pseudorange error is calculated as shown in (5). The other difference is the operating mode of the code tracking loop. In conventional tracking, all code tracking channels are independent closed loops. The feedback to the code NCO is the code discriminator output in each channel. However, in vector tracking, the feedback is calculated using the estimated navigation solution as

235 
$$f_{code,k+1}^{j} = f_{CA} \left[ 1 - \frac{\tilde{\rho}_{k+1}^{j} - \hat{\rho}_{k}^{j}}{c\tau} \right]$$
(18)

where  $\tilde{\rho}_{k+1}^{j}$  and  $\hat{\rho}_{k}^{j}$  are the predicted pseudorange at epoch k+1 and the estimated pseudorange at epoch k. The predicted pseudorange is calculated using

238 
$$\tilde{\rho}_{k+1}^{j} = \left\| \tilde{\mathbf{r}}_{u,k+1} - \mathbf{r}_{k+1}^{j} \right\| + \delta \hat{\rho}_{sv,c}^{j} + \delta \hat{\rho}_{I}^{j} + \delta \hat{\rho}_{T}^{j} - \hat{b}_{clk}$$
(19)

where  $\mathbf{r}_{k+1}^{j}$  and  $\tilde{\mathbf{r}}_{u,k+1}$  are the satellite position and the predicted receiver position at epoch k + 1, respectively.  $\mathbf{r}_{k+1}^{j}$  is known from the broadcast ephemeris, while  $\tilde{\mathbf{r}}_{u,k+1}$  can be calculated based on the estimated position and clock bias at the previous epoch.  $\delta \hat{\rho}_{sv,c}^{j}$ ,  $\delta \hat{\rho}_{I}^{j}$  and  $\delta \hat{\rho}_{T}^{j}$  are the pseudorange errors caused by satellite clock error, ionospheric delay and tropospheric delay, respectively.  $f_{code,k+1}^{j}$  is then fed back to the code NCO in each channel to generate local code replicas.

245

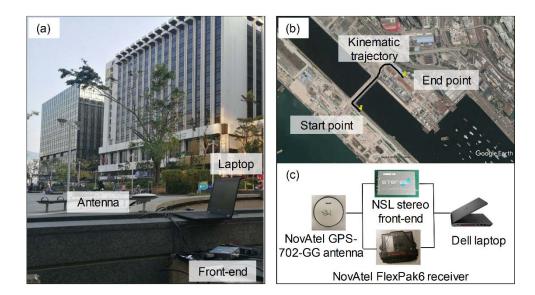
# 246 Vector Tracking

To start vector tracking, initialization parameters, such as ephemeris data, initial receiver PVT, etc., should be provided. The pseudorange error,  $\Delta \rho$ , and pseudo-range rate error,  $\Delta \dot{\rho}$  extracted from the code and carrier tracking loops are used as the measurements of the EKF. The estimated receiver PVT is then used to predict the pseudorange, rate and the LOS vectors at the next epoch.

252

## 253 Experiments and Results

Two experimental tests were conducted to evaluate the performance of vector tracking 254 in terms of its ability against multipath and dynamics effects, respectively. In the first 255 256 test, signals were collected statically in an urban area of Hong Kong, as shown in Fig. 257 3(a). It is expected that the positioning accuracy would decrease due to the potential multipath effects. The second test was conducted in an open-sky environment. In this 258 259 test, the antenna was mounted on the roof of an automobile which kept static for about 30 seconds before moving with a moderate dynamic along a coast, as shown in Fig. 260 261 3(b). A geodetic-grade receiver, NovAtel Flexpak6, was used to provide a reference 262 trajectory. The experimental setup of the kinematic test is shown in Fig. 3(c). In both 263 tests, GPS signals were collected using a Nottingham Scientific Ltd. (NSL) Stereo front-end for post-processing by the developed software. The sampling frequency and 264 265 IF of the front-end are 26 MHz and 6.5 MHz, respectively. In both tests, the update interval of the EKF is one millisecond. The process noise covariance matrix is a 266 diagonal matrix, with its main diagonal values empirically 267 set as diag [0.2, 0.2, 0.2, 0.1, 0.1, 0.1, 0.1, 0.01]. Here, diag [•] denotes a diagonal matrix. The 268 measurement noise is calculated adaptively using equations (15)-(16). 269



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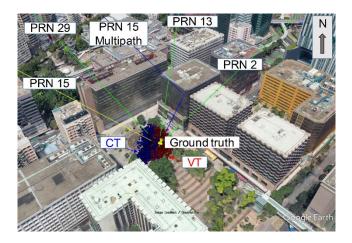


Fig. 3 Experimental environments and setup

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## 273 Static Test Results

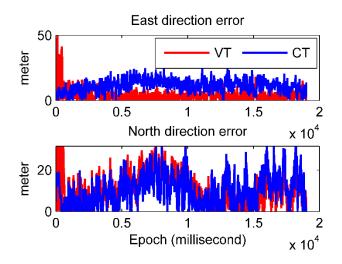
In this test, the receiver antenna was surrounded by high buildings. Only four GPS satellites (PRN 2, 13, 15 and 29) can be acquired and tracked continually using the software receiver, as shown in Fig. 4. Fig. 4 also shows the ray tracing (Hsu et al. 2015a) results of these four satellites based on the ground truth position, among which PRN 15 is a multipath signal, with its direct and reflected signal paths marked in yellow and blue, respectively.



280

Fig. 4 Positioning results and ray tracing results of the four trackable GPS satellites

Fig. 5 presents the positioning errors in east and north directions of vector tracking and conventional tracking during about 20 seconds. The conventional tracking exhibits a mean offset of 11.29 meters in the east direction, while vector tracking remains a lower mean positioning error of 4.19 meters. In north direction, the two methods have similar performance, with a mean error of 10.26 meters and 10.89 meters for vector tracking and conventional tracking, respectively.



289

Fig. 5 Positioning errors in east and north direction.

The positioning offset is probably due to the multipath effect from PRN 15. The 290 291 mechanism by which the vector tracking outperforms the convention tracking in terms 292 of multipath mitigation can be seen in Fig. 6, which demonstrates the code discriminator output and code frequency of PRN 15. Even though the code discriminator output of 293 vector tracking is noisy, the code frequency which directly determines the local code 294 295 replica generation is slightly more stable for vector tracking. This improvement is due 296 to the fact that the code frequency is calculated not only from the measurements but 297 also using the system propagation model. The bottom of Fig. 6 shows the pseudorange measurement variance of PRN 15. Vectoring tracking reports a larger measurement 298 299 variance during the whole test, which indicates that the measurement of PRN 15 contributes less in positioning. 300

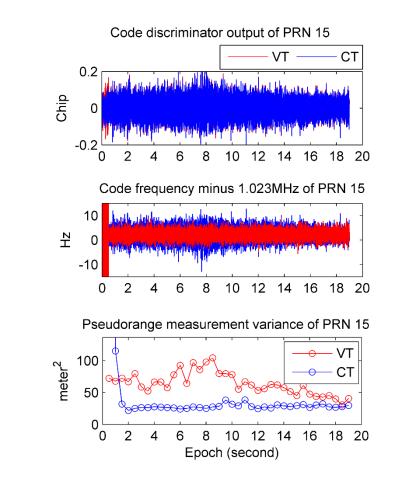


Fig. 6 Code discriminator output, code frequency and pseudorange measurement
 variance of PRN 15

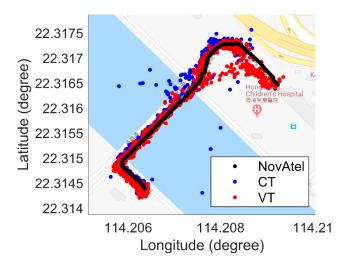
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305 Kinematic Test Results

306 Fig. 7 shows the kinematic positioning results of vector tracking, conventional tracking

307 and NovAtel receiver in a Google map. The U-shape trajectory contains two right turns,

308 a quarter turn and a round turn with a radius of about 40 meters.



310311

#### map

Fig. 7 Positioning results in the kinematic test in an open-sky area plotted in a Google

312 The NovAtel Flexpak6 is a dual-frequency plus L-Band GNSS receiver, thus it has the best positioning result, which is used as the reference for evaluating the other two 313 methods. As seen from Fig. 7, both vector tracking and conventional tracking perform 314 well in the static stage. However, conventional tracking has a large positioning error 315 316 near the round turn. This is due to the signal tracking failure caused by the automobile dynamics, which can be confirmed in Fig. 8. As can be seen in upper panel of Fig. 8, at 317 around 50 seconds, the CNR of PRN 31 suffers a sudden decrease. About 2 seconds 318 later, the value returns to the regular level, which indicates that the tracking loop of 319 320 PRN 31 relocks onto this signal. PRN 12 also suffers from this problem at around 75 seconds (Period B in the vertical yellow shadow), but it takes more time to recover. 321 After that, the CNR values of PRN 25, 21 and 31 decrease successively (Period C in 322 purple shadow). Unfortunately, these tracking loops never relock onto the lost signals. 323 324 Looking at the lower part of Fig. 8, the velocity values have a high correlation with the CNR values, which means the decrease of CNR is caused by the automobile dynamics. 325 The middle panel in Fig. 8 is the CNR of vector tracking. Compared with that of 326 conventional tracking, vector tracking also suffers from the automobile dynamics, but 327 328 after a period of time, the lost signals (PRNs 31, 12, 25 and 21) can be relocked in 329 vector tracking. This is because the code frequency of the lost signal can be predicted using the navigation solution calculated using the information of other channels in 330



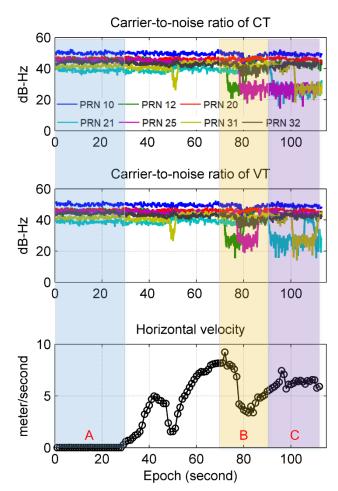
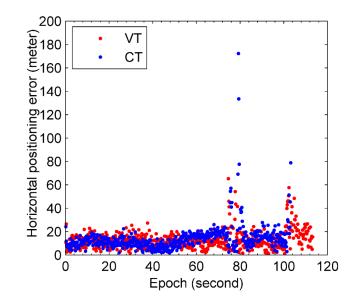


Fig. 8 Carrier-to-noise ratio of vector tracking and conventional tracking, and the
 horizontal velocity during the kinematic test

The horizontal positioning errors of conventional tracking and vector tracking are presented in Fig. 9. The detailed quantitative positioning errors are listed in Table 1. It can be seen that in the static stage, the two methods have similar performances. However, in the kinematic process, vector tracking has a lower positioning error than conventional tracking, especially after 50 seconds when the automobile is in acceleration and deceleration processes.



342 **Fig. 9** Horizontal positioning error of vectoring tracking and conventional tracking.

343 The reference trajectory is provided by NovAtel Flexpak6 receiver

344

345

Table 1 Horizontal positioning errors in three selected periods

Period (second	Error (meter)		
		СТ	VT
Static period	A (1-30)	10.79	11.70
Kinematic	B (70-90)	21.14	14.31
period	C (91-113)	679.89	16.19

346

## 347 Conclusions

A GPS SDR based on vector tracking is implemented in this paper. The algorithm design of vector delay lock loop is presented, with emphasis on the design of the EKF. A conventional tracking-based receiver is also developed, which calculates the receiver navigation solution using the same EKF as vectoring tracking. Static and kinematic tests are conducted in an urban area and an open-sky environment, respectively, to evaluate the performance of vectoring tracking and conventional tracking. Results show that vector tracking has a better capability against signal interference, e.g., multipath signal. Besides, in terms of dynamic performance, vector tracking outperforms conventional tracking due to its coupling of all tracking channels.

The open-source GPS SDR can be used a basic tool to learn the principle of vector tracking and compare its performance with conventional tracking. The contents and functionalities of this software will be continually improved. The current MATLAB software can be found on the GPS Toolbox website at: <u>https://www.ngs.noaa.gov/gps-</u> toolbox. A user manual is also provided, which shows how to install the software and how to process the data collected using a front-end. Any comments, suggestions or corrections are welcome; please send these to the authors.

364

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