

Iterative least squares method for global positioning system

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Abstract. The efficient implementation of positioning algorithms is investigated for Global Positioning System (GPS). In order to do the positioning, the pseudoranges between the receiver and the satellites are required. The most commonly used algorithm for position computation from pseudoranges is non-linear Least Squares (LS) method. Linearization is done to convert the non-linear system of equations into an iterative procedure, which requires the solution of a linear system of equations in each iteration, i.e. linear LS method is applied iteratively. CORDIC-based approximate rotations are used while computing the QR decomposition for solving the LS problem in each iteration. By choosing accuracy of the approximation, e.g. with a chosen number of optimal CORDIC angles per rotation, the LS computation can be simplified. The accuracy of the positioning results is compared for various numbers of required iterations and various approximation accuracies using real GPS data. The results show that very coarse approximations are sufficient for reasonable positioning accuracy. Therefore, the presented method reduces the computational complexity significantly and is highly suited for hardware implementation.

1 Introduction

Location Based Services (LBS) (Monsondo et al., 2006; Perusco, 2002; He and Bilgic, 2009; Schiller and Voisard, 2004) are wireless “mobile content” services which are used to provide location-specific information to mobile users moving from location to location. They utilize the ability to make use of the geographical position of the mobile device. Currently, GPS technique (Schreiner, 2007; Djuknic and Richton, 2002), network positioning methods (Drane et al., 1998) as well as other positioning methods (He et al., 2008; Chan and Ho, 1994) are commonly used positioning methods for location estimation for LBS.

The result of the free availability of satellite positioning parameters has led to wide adoption of the GPS systems. Building GPS devices in commercially available cell phones has been achieved by mobile device providers, such that the number of cell phones equipped with GPS functionality has rapidly grown in the last few years.

In order to do the positioning, an initial set of pseudoranges between the receiver and the satellites is needed. Non-linear LS is the most common method to determine the receiver’s position from the pseudoranges. Usually, linearization is done to convert the non-linear problem into an iterative algorithm, which requires the solution of an overdetermined system of linear equations in each iteration step itr , i.e. linear LS method is applied in each iteration step itr . For solving the linear LS problems in each iteration step an iterative version of the QR decomposition (QRD) (Götze, 1994) is applied in this paper. Instead of annihilating the lower diagonal elements during the QRD, CORDIC-based approximate rotations are used. By choosing the accuracy of the approximation, e.g. by choosing itg optimal CORDIC angles per rotation, the LS computation can be simplified. However, we only obtain an approximate solution to the LS problem, whose accuracy depends on itg . The accuracy of the positioning results of GPS method is compared for varying numbers of iterations itr of the positioning algorithms and varying numbers of iterations itg of the iterative QRD using real GPS data. The results show that very coarse approximations (small itg) are sufficient for obtaining a reasonable position estimate. Therefore the presented methods reduce the computational complexity and the required power consumption significantly.

GPS positioning is introduced in Sect. 2 resulting in an algorithm, which requires the solution of LS problems in each iteration itr ($itr = 1, 2, \dots, itr_{max}$). For solving these LS problems an iteration version of the QRD is presented in Sect. 3 using CORDIC-based approximate rotations, where itg denotes the approximation accuracy ($itg = 1, 2, \dots, itg_{max}$). The trade-off between itr (iteration of the positioning method) and itg (iteration of the iterative QRD) is investigated in Sect. 4, where experimental results are given



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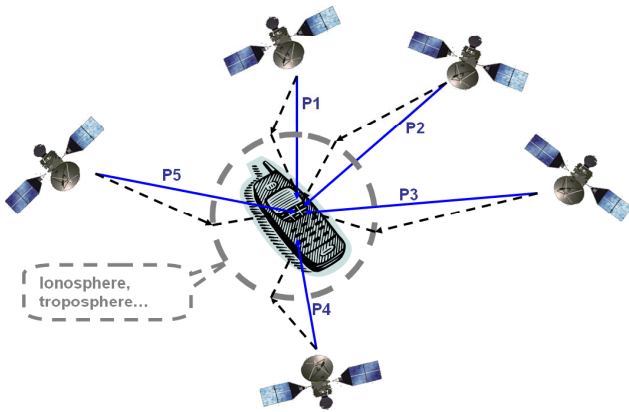


Fig. 1. Pseudoranges: the distance from satellites to GPS receiver.

using real GPS data. The paper finishes with a conclusion and an outlook to the future work in Sect. 5.

2 GPS positioning

The whole GPS positioning procedure includes three tasks: acquisition, tracking and positioning (Borre et al., 2007). The acquisition tries to find satellites and to get their positions. It gives rough estimates of signal parameters. Tracking keeps track of these parameters as the signal properties change over time. After tracking, the navigation data can be extracted and pseudoranges (measured distance from satellites to GPS receiver) can be computed. The final task of the receiver is to compute the user position.

The GPS satellites' arrangement ensures that every point on our planet is in contact with at least six satellites at all times. Each satellite k continuously broadcasts a digital radio signal that includes its position (X^k, Y^k, Z^k) and its time t^k . On board atomic clocks ensure an accurate time to a billionth of a second. The radio signal of satellite spreads with $c = 3 \times 10^8$ in universe, the velocity of light in vacuum. GPS receivers measure the time delay τ^k of the signal from each satellite k to the receiver, so $\tau^k = t - t^k$, where t is time of receiver. The measurement of t in the receiver is not very accurate (as compared to the satellite time t^k). Furthermore the speed of radio signal from the satellites cuts down because of ionosphere and troposphere. Signal propagation duration from satellites to the receiver is longer than expected. Therefore the measured distance from the satellite to the receiver P^k , measured by $P^k = \tau^k \cdot c$, is a rough distance estimate called "Pseudorange" (see Fig. 1). The receiver simultaneously collects these measurements from at least four satellites and processes them to solve for position and time measurement error.

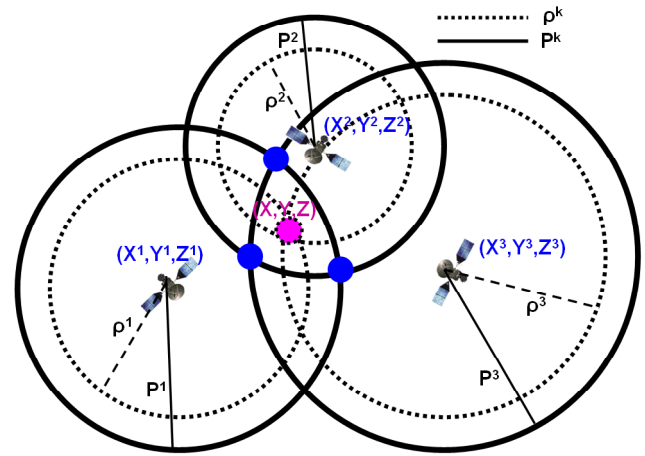


Fig. 2. Observed pseudoranges P^k and geometrical pseudoranges ρ^k .

2.1 Observation equation

The most commonly used algorithm for position computation from pseudoranges is based on the LS method.

This method is used to find the receiver position from pseudoranges to four or more satellites.

The basic observation equation for the pseudorange P^k is

$$P^k = \rho^k + c(dt - dt^k) + T^k + \ell^k + e^k. \quad (1)$$

ρ^k is the geometrical range between satellite k and receiver, which can be computed as:

$$\rho^k = \sqrt{(X^k - X)^2 + (Y^k - Y)^2 + (Z^k - Z)^2} \quad (2)$$

where (X, Y, Z) is the position of receiver (see Fig. 2). dt denotes the receiver clock offset and dt^k is the satellite clock offset. From the ephemerids, which also include information on the satellite clock offset dt^k , the position of the satellite (X^k, Y^k, Z^k) can be computed. T^k is the tropospheric error and ℓ^k is the ionospheric error. These two errors are computed from a priori models, whose coefficients are part of the broadcast ephemerids. e^k is the observation error of the pseudorange. Therefore, Eq. (1) contains four unknowns X, Y, Z and dt . The error terms are minimized by using the LS method.

Equation (2) is nonlinear with respect to the receiver position (X, Y, Z) , so the equation is linearized before using the LS method. The nonlinear term in Eq. (2):

$$f(X, Y, Z) = \sqrt{(X^k - X)^2 + (Y^k - Y)^2 + (Z^k - Z)^2} \quad (3)$$

is linearized. Starting from an initial position for the receiver (X_1, Y_1, Z_1) , the position estimate is improved iteratively. The center of the Earth $(0, 0, 0)$ can be chosen as the initial point, if no a-priori-information (as e.g. from AGPS) is available. Let itr be the number of the iterations

($itr = 1, 2, \dots, itr_{\max}$). The increments $\Delta X_{itr}, \Delta Y_{itr}, \Delta Z_{itr}$ update the receiver coordinates as follows:

$$\begin{aligned} X_{itr+1} &= X_{itr} + \Delta X_{itr}, \\ Y_{itr+1} &= Y_{itr} + \Delta Y_{itr}, \\ Z_{itr+1} &= Z_{itr} + \Delta Z_{itr}. \end{aligned} \quad (4)$$

The Taylor expansion of $f(X_{itr} + \Delta X_{itr}, Y_{itr} + \Delta Y_{itr}, Z_{itr} + \Delta Z_{itr})$ is

$$\begin{aligned} f(X_{itr+1}, Y_{itr+1}, Z_{itr+1}) &= f(X_{itr}, Y_{itr}, Z_{itr}) \\ &+ \frac{\partial f(X_{itr}, Y_{itr}, Z_{itr})}{\partial X_{itr}} \Delta X_{itr} \\ &+ \frac{\partial f(X_{itr}, Y_{itr}, Z_{itr})}{\partial Y_{itr}} \Delta Y_{itr} \\ &+ \frac{\partial f(X_{itr}, Y_{itr}, Z_{itr})}{\partial Z_{itr}} \Delta Z_{itr} \end{aligned} \quad (5)$$

Equation (5) includes only first-order terms, and the partial derivatives are

$$\begin{aligned} \frac{\partial f(X_{itr}, Y_{itr}, Z_{itr})}{\partial X_{itr}} &= -\frac{X^k - X_{itr}}{\rho_{itr}^k} \\ \frac{\partial f(X_{itr}, Y_{itr}, Z_{itr})}{\partial Y_{itr}} &= -\frac{Y^k - Y_{itr}}{\rho_{itr}^k} \\ \frac{\partial f(X_{itr}, Y_{itr}, Z_{itr})}{\partial Z_{itr}} &= -\frac{Z^k - Z_{itr}}{\rho_{itr}^k} \end{aligned}$$

Let $\rho_{itr}^k = \sqrt{(X^k - X_{itr})^2 + (Y^k - Y_{itr})^2 + (Z^k - Z_{itr})^2}$ be the range computed from the satellite position (X^k, Y^k, Z^k) to the approximate receiver position $(X_{itr}, Y_{itr}, Z_{itr})$, so the first-order linearized observation equation becomes

$$\begin{aligned} P_{itr}^k &= \rho_{itr}^k - \frac{X^k - X_{itr}}{\rho_{itr}^k} \Delta X_{itr} - \frac{Y^k - Y_{itr}}{\rho_{itr}^k} \Delta Y_{itr} \\ &- \frac{Z^k - Z_{itr}}{\rho_{itr}^k} \Delta Z_{itr} + c(dt_{itr} - dt^k) + T_{itr}^k + \ell_{itr}^k + e_{itr}^k. \end{aligned} \quad (6)$$

where dt_{itr} is the estimated clock error at the receiver.

2.2 Applying least-squares method

Rearranging Eq. (6), it can be rewritten as

$$\begin{aligned} &\left[-\frac{X^k - X_{itr}}{\rho_{itr}^k} - \frac{Y^k - Y_{itr}}{\rho_{itr}^k} - \frac{Z^k - Z_{itr}}{\rho_{itr}^k} + 1 \right] \begin{bmatrix} \Delta X_{itr} \\ \Delta Y_{itr} \\ \Delta Z_{itr} \\ cdt_i \end{bmatrix} \\ &= P_{itr}^k - \rho_{itr}^k + cdt^k - T_{itr}^k - \ell_{itr}^k - e_{itr}^k. \end{aligned} \quad (7)$$

A unique solution can not be found from a single equation. Therefore $m \geq 4$ satellites are required to form a system of linear equations (usually $m \geq 6$ satellites are available). Let $b_{itr}^k = P_{itr}^k - \rho_{itr}^k + cdt^k - T_{itr}^k - \ell_{itr}^k - e_{itr}^k$ and

$\mathbf{b}_{itr} = [b_{itr}^1, b_{itr}^2, \dots, b_{itr}^m]^T$. Then we obtain the LS problem:

$$\min_{\mathbf{x}_{itr}} \|\mathbf{A}_{itr} \mathbf{x}_{itr} - \mathbf{b}_{itr}\|_2, \quad (8)$$

where

$$\mathbf{A}_{itr} = \begin{bmatrix} -\frac{X^1 - X_{itr}}{\rho_{itr}^1} & -\frac{Y^1 - Y_{itr}}{\rho_{itr}^1} & -\frac{Z^1 - Z_{itr}}{\rho_{itr}^1} & 1 \\ -\frac{X^2 - X_{itr}}{\rho_{itr}^2} & -\frac{Y^2 - Y_{itr}}{\rho_{itr}^2} & -\frac{Z^2 - Z_{itr}}{\rho_{itr}^2} & 1 \\ -\frac{X^3 - X_{itr}}{\rho_{itr}^3} & -\frac{Y^3 - Y_{itr}}{\rho_{itr}^3} & -\frac{Z^3 - Z_{itr}}{\rho_{itr}^3} & 1 \\ \vdots & \vdots & \vdots & \vdots \\ -\frac{X^m - X_{itr}}{\rho_{itr}^m} & -\frac{Y^m - Y_{itr}}{\rho_{itr}^m} & -\frac{Z^m - Z_{itr}}{\rho_{itr}^m} & 1 \end{bmatrix}$$

and $\mathbf{x}_{itr} = [\Delta X_{itr} \ \Delta Y_{itr} \ \Delta Z_{itr} \ cdt_{itr}]^T$. The solution $\Delta X_{itr}, \Delta Y_{itr}, \Delta Z_{itr}$ has to be added to the approximate receiver position to get the next approximate position as in Eq. (4). These iterations continue until the solution $\Delta X_{itr}, \Delta Y_{itr}, \Delta Z_{itr}$ is at meter level.

For solving Eq. (8) we need to find $\hat{\mathbf{x}}_{itr}$ which minimizes the length of the error vector $\hat{\mathbf{e}}_{itr} = \mathbf{b}_{itr} - \mathbf{A}_{itr} \hat{\mathbf{x}}_{itr}$ with $\|\hat{\mathbf{e}}_{itr}\|^2 = (\mathbf{b}_{itr} - \mathbf{A}_{itr} \hat{\mathbf{x}}_{itr})^T (\mathbf{b}_{itr} - \mathbf{A}_{itr} \hat{\mathbf{x}}_{itr})$ the sum of squares of the m separate errors. Minimizing this quadratic gives the normal equations

$$\mathbf{A}_{itr}^T \mathbf{A}_{itr} \hat{\mathbf{x}}_{itr} = \mathbf{A}_{itr}^T \mathbf{b}_{itr} \Rightarrow \hat{\mathbf{x}}_{itr} = (\mathbf{A}_{itr}^T \mathbf{A}_{itr})^{-1} \mathbf{A}_{itr}^T \mathbf{b}_{itr} \quad (9)$$

and the error vector is

$$\hat{\mathbf{e}}_{itr} = \mathbf{b}_{itr} - \mathbf{A}_{itr} \hat{\mathbf{x}}_{itr}. \quad (10)$$

There are various algorithms for solving LS problems (Eq. 8) (Golub and Loan, 1996). In the subsequent section, we use the QRD by Givens rotations.

3 Iterative solution of LS

In each iteration of the positioning algorithm a LS problem must be computed. Since the pseudoranges are subject to measurement errors and the convergence (number of required iterations) of the algorithm depends on the accuracy of these LS solutions, it is worthwhile to investigate the use of an iterative LS solver and the trade-off between the number of iterations of the positioning method and the number of iterations of the iterative LS solver.

The iterative version of the QR decomposition (QRD) presented in Götze (1994) is used for the iterative solution of Eq. (8), since it is well suited for hardware implementation, suitable for the adaption to measurement errors (pseudoranges), and yields a solution vector which converges linearly to the exact solution. Here we will briefly review this iterative QRD.

The QR decomposition of a $m \times n$ matrix $\mathbf{A} = \mathbf{Q} \cdot \mathbf{R}$ can be computed by applying a sequence of Givens rotations $\mathbf{G}(\phi_{ij})$

Table 1. Mean value of position-accuracy results (in meter) by GPS with exact LS method and CORDIC-based approximate rotations.

<i>itr</i>	<i>itg</i>	GPS-itr	GPS-itg
2	1		1282.8
	2		1256.6
	3	1241.4	1246.1
	8		1241.4
3	9		1241.4
	1		85.756
	2		47.744
	3	38.484	39.020
4	8		38.484
	9		38.484
	1		81.495
	2		35.452
5	3	32.472	32.985
	8		32.472
	9		32.472
	1		81.495
5	2		35.452
	3	32.472	32.985
	8		32.472
	9		32.472

($\phi_{ij} = \arctan(a_{ij}/a_{jj})$) to the matrix such that the matrix entries below the diagonal of \mathbf{A} are annihilated, i.e. generate $a'_{ij} = 0$ for

$$(i, j) = \{(2, 1)(3, 1) \dots (m, 1)(3, 2) \dots (m, 2) \dots (n+1, n) \dots (m, n)\}. \quad (11)$$

The resulting matrix is upper triangular and denoted by \mathbf{R} . The product of all required Givens rotations forms the orthogonal matrix $\mathbf{Q}^T = \prod_{j=1}^n \prod_{i=j+1}^m \mathbf{G}(\phi_{ij})$ such that

$$\min_{\mathbf{w}} \|\mathbf{A}\mathbf{w} - \mathbf{b}\|_2 = \min_{\mathbf{w}} \|\mathbf{R}\mathbf{w} - \mathbf{Q}^T \mathbf{b}\|_2 \quad (12)$$

and the solution \mathbf{w} can be computed by back substitution.

One iteration of the iterative version of the QRD works exactly like the QRD but instead of using exact rotations $\mathbf{G}(\phi_{ij})$ that annihilate a_{ij} ($a'_{ij} = 0$) CORDIC-based approximate rotations are used resulting in $|a'_{ij}| = |d||a_{ij}|$ with $0 \leq |d| < 1$. The CORDIC-based approximate Givens rotations are computed by determining the shift value ℓ , $\ell \in \{0, 1, 2, \dots, b\}$ (b wordlength) corresponding to the CORDIC angle $\phi_{ij}(\ell) = \arctan 2^{-\ell}$ which is closest to the exact rotation angle ϕ_{ij} . Instead of annihilating the matrix elements during the course of the QRD an approximate rotation will only reduce the matrix elements by the maximal factor possible with itg specific CORDIC angles $\phi_{ij}(\ell)$. This CORDIC-based approximate rotation is applied to the QRD for solving LS problems.

Since the matrix elements below the diagonal are no longer annihilated the QRD procedure using CORDIC-based approximate rotations must iteratively be applied until the matrix is ultimately upper triangular. One obtains *iterative* versions of the QRD distinguished by itg , which determines the accuracy of the approximation of the rotations.

Defining the lower diagonal quantity for iteration itg :

$$S^{(itg)} = \sum_{j=1}^n \sum_{i=j+1}^m \left(a_{ij}^{(itg)} \right)^2 \quad (13)$$

the above algorithm guarantees

$$\lim_{itg \rightarrow \infty} S^{(itg)} \rightarrow 0$$

which is equivalent to

$$\lim_{itg \rightarrow \infty} \mathbf{A}^{(itg)} \rightarrow \mathbf{R} \quad \text{and} \quad \lim_{itg \rightarrow \infty} \mathbf{Q}^{T(itg)} \rightarrow \mathbf{Q}^T$$

When we apply this method to the LS problems, which must be solved for positioning in Eq. (8), very few steps, i.e. $itg \ll b$, of the CORDIC-based approximate rotation are sufficient to obtain similar results as using exact rotations. This yields a significant reduction in computational complexity by itg/b (we will show in the following section for real GPS data that even $itg = 1$ gives reasonable results). Furthermore, this method only requiring shift and add operations is very well suited for hardware implementation.

4 Experimental results

4.1 GPS raw data collection

For GPS positioning, a SiGe GN3S Sampler v2 is used to capture the raw GPS data, which are low level signal data (raw intermediate frequency samples) being delivered by the GPS satellite network and processed by the SiGe radio front end (GN3S, 2009). Each GPS satellite continuously broadcasts a navigation message at a rate of 50 bits per second.

The obtained raw GPS data at each measurement point includes information of $m = 7$ satellites with matched pseudoranges. 78 raw GPS data records are gained. Afterwards position calculation is done for GPS method for various numbers of required iterations itr (see Sect. 2) and various approximation accuracies itg (see Sect. 3).

4.2 Experimental results

The mean values of the positioning results for the 78 measurements are presented in Table 1. The calculated positions are compared with the exact positions (known coordinates at the measure points from land surveying office in Bochum Germany) and the accuracy of the positioning results of GPS method for varying itr and itg is compared. The number of required iterations itr is at least two. The second column itg

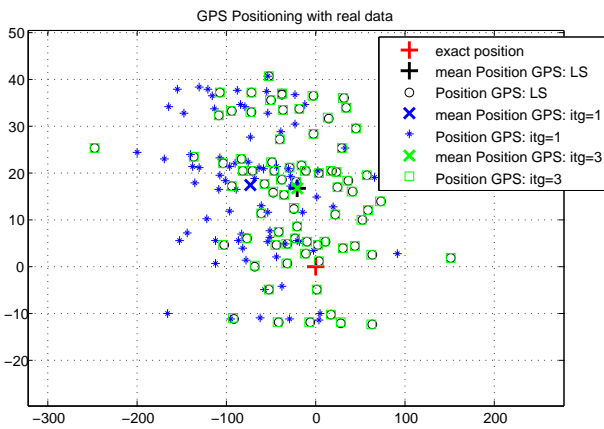


Fig. 3. Positioning results with GPS method.

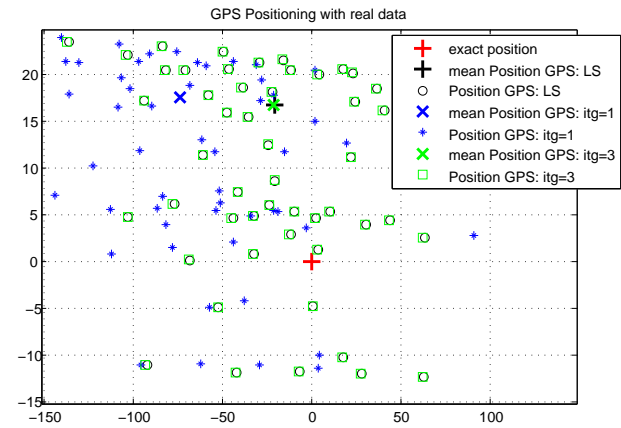


Fig. 4. Enlarged positioning results with GPS method.

in Table 1 is the approximation accuracy in QRD. Figures 3 and 4 show the position estimates of GPS positioning. In Table 1 the 3rd column GPS-itr is the accuracy of GPS positioning with exact LS method and the 4th column GPS-itg is the accuracy of GPS positioning using QRD with CORDIC-based approximate rotations for solving LS problems. The results show that with the increasing number of iterations in QRD, the accuracy of GPS-itg also increases. If $itg \geq 8$, the accuracy of the exact LS method is achieved. Especially, if itr is big enough, e.g. $itr \geq 3$, only one iteration in QRD $itg = 1$ is required to compute the position with reasonable positioning accuracy, which means very coarse approximations are sufficient.

The position estimates of 78 measurements are shown in Fig. 3. GPS position with exact LS (black circles), its mean value (black cross), GPS position using iterative QRD with $itg = 1$ (blue stars), its mean value (blue cross), GPS position using iterative QRD with $itg = 3$ (green squares), its mean value (green cross) and the exact position (red cross) are shown in the figure. All the position results are considered as accuracy of position, i.e. the position estimates subtracted by the exact position (so the exact position is set to (0, 0)).

Figure 4 is an enlarged part of Fig. 3. It is obvious to notice that GPS position using QRD with $itg = 3$ (the green squares) are more closer to GPS position with exact LS (black circles) than GPS position using QRD with $itg = 1$ (blue stars). If the number of iterations in QRD increases, i.e. itg increases, the position results will become more and more similar to the results of exact LS. The mean value of $itg = 3$ is almost the same as the mean value of exact LS (green cross is almost at the same position as black cross in Fig. 4).

5 Conclusion

An iterative LS approach and an iterative version of the QRD using CORDIC-based approximate rotation are applied to position computation. The accuracy of the positioning re-

sults of GPS method is compared for various numbers of required iterations itr and various approximation accuracies itg of CORDIC-based approximate rotations by using real GPS data. It is shown that a significant reduction concerning computational complexity and hardware requirements can be obtained. Furthermore, this method only requiring shift and add operations is very well suited for hardware implementation.

The presented method is very efficient for the implementation of the standard triangulation method based on non-linear LS. In future work we apply this idea to recursive computations of the position estimates using Orthogonal DGPS (Chang and Paige, 2003) and Kalman Filter based recursive GPS algorithms (Grewall et al., 2001). In the first case no iterative LS method is required for positioning which leads to further reduction of the computational effort and the power consumption. In the second case using the square root version of the Kalman filter (Merwe and Wan, 2001) QRD can be applied and the required approximation accuracy (number of itg) can be investigated to obtain desired position accuracy.

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