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CONSIDERING VARIANCES OF QUASI-RANDOM EFFECTS IN RELATIVE GPS POSITIONING PERFORMED DURING DAYTIME AND NIGHTTIME PERIODS – A NOVEL TWO-STAGE APPROACH

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Abstract. In this paper, a new two-stage approach, involving an integral treatement of all quasi-random effects limiting the accuracy of relative GPS positioning and the method of moments to obtain final variance components regarding the effects of short-term ("far-field") multipath (factor β), joint action of long-term ("near-field") multipath and receiver antenna phase center offset and variations (factor α 1), as well as joint action of tropospheric and ionospheric refraction (factor α 2), is presented. In the study, GPS data collected on five baselines were used. Variance components of the quasirandom effects were obtained for the three relative GPS coordinates (*e*, *n* and *u*) using individually monthly datsets including daytime- and those including nighttime-wise ambiguity-fixed baseline solutions. The related results show that statistically significant inequality exists when comparing corresponding variances obtained for daytime and nighttime periods. It turned out that the following standard deviation estimates intervals are present (by the coordinates *e*, *n* and *u*); 0.0116– 0.3282, 0.0103–0.2365 and 0.1222–0.7818 mm/km (factor α 2); (b) nighttime period: 3.2–4.9, 4.7–7.3 and 8.4–15.4 mm (factor β); 0.8–3.8, 2.1–5.0 and 3.1–15.8 mm (factor α 1); 0.0118–0.2734, 0.0097–0.2289 and 0.0752–0.6315 mm/km (factor α 2).

Keywords: relative GPS positioning, quasi-random effects, variance components, ANOVA estimates, method of moments, sub-daily impacts, statistical hypotheses testing.

Introduction

Many studies have been conducted with an aim to consider and represent statistical behaviour of unmodelled, i.e. residual effects arising in the relative GPS positioning. But, in each of them these effects were treated individually and a reader can be referred to the recent ones, such as, e.g. Hu et al. (2015), Deng et al. (2016), Elsobeiey and El-Diasty (2016), Jadviščok et al. (2016), Klos et al. (2018), Zhou et al. (2018), Anđić, (2019b), Han et al. (2019), Juni and Rózsa (2019), Kallio et al. (2019).

Herein, however, the author uses a linear model involving effects limiting relative GPS positioning accuarcy at once. Namely, the effects, such as short- ("far-field" multipath – FF MP) and long-term ("near-field" multipath – NF MP, receiver antenna phase center offset and variations – RAPCOV, tropospheric and ionospheric refraction – TI) unmodelled effects, are the only ones that must not be neglected when making a choice of baselines as in this study (each of them is of less than 300 km in length). So, in this regard, only the previously mentioned effects are being considered in this study.

In the previous work, related to the use of the twoway nested classification in GPS positioning, published by Anđić (2016, 2019a), only a baseline of 40 km in length was considered, and only ANOVA estimates obtained on the basis of daily datasets were presented. Herein, a further analysis, dealing with ANOVA estimates obtained by using sub-daily (daytime and nighttime) datasets is performed. In addition, one more method (method of moments, as a tool which the second stage of the approach applied here is based on), as well as GPS data collected at ten ending stations two of which belong to MontePos (Montenegrin Permanent Network) (PODG and BAR), and remaining eight that have been integrated in EPN (EUREF Permanent Network) infrastructure (TORI, IENG, IGMI, PRAT, BOR1, WROC, BACA and BAIA). These stations establish five baselines, ranging from 5.6 to 281.9 km in length. All of the ten stations have been equipped with high-quality antennas, mounted in different way and surrounded with

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This is an Open Access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/4.0/), which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited. potential reflectors of different characteristics, some of which were in the near vicinity while others were far away from an antenna. More detailed information related to the two MontePos stations can be found at the following link: http://www.nekretnine.co.me/me/Djelatnosti5.asp. On the other side, information regarding eight EPN stations are provided at the following address: http://epncb.oma. be/_networkdata/stationmaps.php.

It should be noted that no paper dealing with the previously mentioned effects in the way presented in this study has been published so far. That is where the motivation for writing this paper came from.

1. Data and methods

1.1. Input data

For the purpose of the study, a total number of 38548930 true errors (corresponding to the relative coordinates *e*, *n* and *u*) remained after outliers removal, were used. These solutions were obtained by using Trimble Total Control software in processing 0.033-Hz GPS data, collected, as previously said, at ending stations of five baselines. The baselines are located in Italy (TORI-IENG and IGMI-PRAT, being 5.6 and 13.6 km long, respectively), in Montenegro (BAR-PODG, 40 km long), in Poland (BOR1-WROC, 129.5 km long) and in Romania (BACA-BAIA, of 281.9 km in length).

In data processing, the processing interval of 30 s, precise orbits, elevation cutoff of 10°, ionosphere-free linear combination, default calibration model, Saastamoinen tropospheric model, MSIS meteorological model and OTF processing mode (as an auxiliary one) were used. Such settings allowed all unmodelled effects to be expressed in each ambiguity-fixed solution.

1.2. The two-stage method used in the study

In the first stage of the study, the two-way nested ANOVA was applied individually on monthly datasets of daytimeand nighttime-wise true errors for all of the three relative coordinates (*e*, *n* and *u*), whereby variance components estimates $m_{\varepsilon,c}^2 = \widehat{\sigma_{\varepsilon,c}^2}$, $m_{\beta,c}^2 = \widehat{\sigma_{\beta,c}^2}$ and $m_{\alpha,c}^2 = \widehat{\sigma_{\alpha,c}^2}$, with $c \in \{e,n,u\}$, were calculated. Herein, the following denotations were introduced: α – the nesting factor (related to the unmodelled joint long-term effect, including NF MP, RAPCOV, TI); β – *the nested factor, nested within* α (related to the unmodelled short-term effect, i.e. FF MP); ϵ – *the purely random error, nested within* β . It should also be mentioned that the 3-minute constancy of the short-term effect (i.e. FF MP – the nested factor β) and the 90-minute constancy of all long-term effects (i.e. NF MP, RAPCOV, TI – the nesting factor α) were assumed in the calculations.

The concept of the first-stage-related method is shown in Figure 1.

Namely, here we introduce *the true error model equation* of an individual coordinate *c* as follows:

$$\Delta_{c,ijk} = \alpha_{c,i} + \beta_{c,j(i)} + \varepsilon_{c,k(ij)}, \qquad (1)$$

whereby *the accompanying stochastic model* is based on the assumptions below:

$$\alpha_{c,i} \sim N(0, \sigma_{\alpha,c}^2)$$
, with $i \in \{1, 2, ..., a\}$, $a \ge 2$, (2a)

$$\beta_{c,j(i)} \sim N(0,\sigma_{\beta,c}^2)$$
, with $j \in \{1,2,\dots,b_i\}$, $b_i \ge 2$, (2b)

$$\begin{aligned} & \varepsilon_{c,k(ij)} \sim N(0,\sigma_{\varepsilon,c}^2) \text{, with } k \in \{1,2,\ldots,n_{ij}\} \text{, } n_{ij} \ge 2, (2c) \\ & \text{Cov}(\varepsilon_{c,k(ij)},\beta_{c,j(i)}) = \text{Cov}(\varepsilon_{c,k(ij)},\alpha_{c,i}) = \end{aligned}$$

$$\operatorname{Cov}(\beta_{c,j(i)},\alpha_{c,i}) = 0, \qquad (2d)$$

 $\operatorname{Cov}(\varepsilon_{c,k(ij)},\varepsilon_{c,r(pq)})=0$, where $i \neq p \lor j \neq q \lor k \neq r$. (2e)

On the basis of (1) and (2a–e), one writes:

$$\operatorname{Var}(\Delta_{c,ijk}) \equiv \sigma_{\Delta}^2 = \sigma_{\alpha,c}^2 + \sigma_{\beta,c}^2 + \sigma_{\varepsilon,c}^2, \qquad (3)$$

whereby $\sigma_{\alpha,c}^2$, $\sigma_{\beta,c}^2$ and $\sigma_{\epsilon,c}^2$ represent *the variance components that are to be estimated* in the ANOVA estimation procedure. With of an aim to avoid unnecessary transcription of the corresponding formulas, that procedure is not shown herein, so, for details, a reader is reffered to Anđić (2016).

In the second stage, with the aim of calculation of final variance components, an iterative method of moments was used. Namelly, as assumptions regarding unmodelled and purely random effects are $(i \in \{1, 2, ..., a\}, j \in \{1, 2, ..., b_i\}, k \in \{1, 2, ..., n_{ii}\}$):

$$\{\alpha_{c,i} \quad \beta_{c,j(i)} \quad \varepsilon_{c,k(ij)}\}^{\mathrm{T}} \sim N(\mathbf{0}_{3\times 1}, \operatorname{diag}\{\sigma_{\alpha,c}^{2} \quad \sigma_{\beta,c}^{2} \quad \sigma_{\varepsilon,c}^{2}\}), (4)$$



Figure 1. The concept of the method applied herein (adapted from Figure 6.3.1 in Anđić, 2019b)

then one can establish *the second and the fourth* $\xi_{c,s}$ *moment* for those effects as follows ($\xi \in {\alpha, \beta, \varepsilon}$, $s \in {1, 2, ..., g}$, $g \ge k$):

$$M\{\xi_{c,s}^2\} = \sigma_{\xi,c}^2 = \sigma_{l;\xi,c}^2 Q_{l,c,ss} + \dots + \sigma_{k;\xi,c}^2 Q_{k,c,ss};$$
(5)

$$D\{\xi_{c,s}^2\} = \sigma_{\xi_{c,s}^2}^2 = 2\sigma_{\xi,c;s}^4,$$
(6)

where $\sigma_{1;\xi,c}^2,...,\sigma_{k;\xi,c}^2$ are the unknown variance components that are estimated.

For that purpose in this study, considering, in addition, that $M{\xi_{c,s}^2}=M{m_{\xi,c;s}^2}$, one establishes *the linear* and stochastic model of the standard least squares as follows:

$$v_{\alpha,c;s} = \sigma_{1;\alpha,c}^2 + \sigma_{2;\alpha,c}^2 D_{s[km]}^2 - m_{\alpha,c;s}^2;$$
(7a)

$$v_{\beta,c;s} = \sigma_{1;\beta,c}^2 - m_{\beta,c;s}^2;$$
(7b)

$$v_{\varepsilon,c;s} = \sigma_{1;\varepsilon,c}^2 - m_{\varepsilon,c;s}^2; \tag{7c}$$

$$M\{\nu_{\alpha,c;s}\}=0, \quad P_{\alpha,c;s}=N_{\alpha,c;s} / (2\sigma_{\alpha,c;s}^4); \quad (8a)$$

$$M\{\nu_{\beta,c;s}\}=0, \quad P_{\beta,c;s}=N_{\beta,c;s} / (2\sigma_{\beta,c;s}^{4});$$
(8b)

$$\mathbf{M}\{\nu_{\varepsilon,c;s}\}=0, \quad P_{\varepsilon,c;s}=N_{\varepsilon,c;s} / (2\sigma_{\varepsilon,c;s}^4), \quad (8c)$$

where $v_{\xi,c;s}$, $P_{\xi,c;s}$, $N_{\xi,c;s}$ and D_s ($\xi \in \{\alpha,\beta,\epsilon\}$, $c \in \{e,n,u\}$, $s \in \{1,2,3,4,5\}$) are denotations for *corrections*, weights, numbers of daytime- or nighttime-wise true errors within a monthly dataset based on which ANOVA estimates $m_{\xi,c;s}^2$ were obtained, and *baseline lengths*, respectively.

The unknown variance components $\sigma_{1;\alpha,c}^2$, $\sigma_{2;\alpha,c}^2$, $\sigma_{1;\beta,c}^2$ and $\sigma_{1;\epsilon,c}^2$, regarding, respectively, the joint effect of NF MP and RAPCOV, the joint effect TI, the effect of FF MP and the purely random error, are estimated through the following iterative procedure:

Step 1. Adopting initial weight matrices:

$$\mathbf{P}_{\alpha,c}^{(0)} = \mathbf{P}_{\beta,c}^{(0)} = \mathbf{P}_{\varepsilon,c}^{(0)} \equiv \mathbf{I}_{5\times 5} .$$
(9)

Step 2. Calculating initial estimates of variance components:

$$\begin{bmatrix} \widehat{\boldsymbol{\sigma}_{1;\alpha,c}^{2(0)}} \\ \widehat{\boldsymbol{\sigma}_{2;\alpha,c}^{2(0)}} \end{bmatrix} \equiv \begin{bmatrix} m_{1;\alpha,c}^{2(0)} \\ m_{2;\alpha,c}^{2(0)} \end{bmatrix} = (\mathbf{A}^{\mathrm{T}} \mathbf{P}_{\alpha,c}^{(0)} \mathbf{A})^{-1} \mathbf{A}^{\mathrm{T}} \mathbf{P}_{\alpha,c}^{(0)} \mathbf{I}_{\alpha,c}; \quad (10)$$

$$\widehat{\sigma_{1;\beta,c}^{2(0)}} \equiv m_{1;\beta,c}^{2(0)} = (\mathbf{1}_{5\times 1}^{\mathrm{T}} \mathbf{P}_{\beta,c}^{(0)} \mathbf{1}_{5\times 1})^{-1} \mathbf{1}_{5\times 1}^{\mathrm{T}} \mathbf{P}_{\beta,c}^{(0)} \mathbf{1}_{\beta,c}; \qquad (11)$$

$$\widehat{\sigma_{1;\varepsilon,c}^{2(0)}} \equiv m_{1;\varepsilon,c}^{2(0)} = (\mathbf{1}_{5\times 1}^{\mathrm{T}} \mathbf{P}_{\varepsilon,c}^{(0)} \mathbf{1}_{5\times 1})^{-1} \mathbf{1}_{5\times 1}^{\mathrm{T}} \mathbf{P}_{\varepsilon,c}^{(0)} \mathbf{1}_{\varepsilon,c}, \qquad (12)$$

whereby, besides the vector of ones $(\mathbf{1}_{5\times 1})$, we have the following matrix and vectors $(\xi \in \{\alpha, \beta, \epsilon\})$:

$$\mathbf{A} = [\mathbf{1}_{5\times 1} \ \mathbf{d}]; \tag{13}$$

$$\mathbf{d} = \{ D_{1 \ [km]}^2 \ D_{2 \ [km]}^2 \ D_{3 \ [km]}^2 \ D_{4 \ [km]}^2 \ D_{5 \ [km]}^2 \}^{\mathrm{T}}; \quad (14)$$

$$\mathbf{l}_{\xi,c} = \{m_{\xi,c;1}^2 \ m_{\xi,c;2}^2 \ m_{\xi,c;3}^2 \ m_{\xi,c;4}^2 \ m_{\xi,c;5}^2\}^{\mathrm{T}}.$$
 (15)

Step 3. Calculating weight matrices using variance components estimates obtained in Step 2:

$$\mathbf{P}_{\alpha,c}^{(1)} = \text{diag}\{N_{\alpha,c;s} / (m_{1;\alpha,c}^{2(0)} + m_{2;\alpha,c}^{2(0)}D_{s\ [km]}^{2})^{2}\}; \quad (16)$$

$$\mathbf{P}_{\beta,c}^{(1)} = \text{diag}\{N_{\beta,c;s} \ / \ m_{1;\beta,c}^{4 \ (0)}\};$$
(17)

$$\mathbf{P}_{\varepsilon,c}^{(1)} = \text{diag}\{N_{\varepsilon,c;s} \ / m_{1;\varepsilon,c}^{4\ (0)}\}.$$
(18)

Step 4. Calculating new (final) estimates of variance components:

$$\begin{bmatrix} \widehat{\sigma_{1;\alpha,c}^{2(1)}} \\ \widehat{\sigma_{2;\alpha,c}^{2(1)}} \end{bmatrix} \equiv \begin{bmatrix} m_{1;\alpha,c}^{2(1)} \\ m_{2;\alpha,c}^{2(1)} \end{bmatrix} = (\mathbf{A}^{\mathrm{T}} \mathbf{P}_{\alpha,c}^{(1)} \mathbf{A})^{-1} \mathbf{A}^{\mathrm{T}} \mathbf{P}_{\alpha,c}^{(1)} \mathbf{I}_{\alpha,c}; \quad (19)$$

$$\widehat{\mathbf{p}}_{1;\beta,c}^{2(1)} \equiv m_{1;\beta,c}^{2(1)} = (\mathbf{1}_{5\times 1}^{\mathrm{T}} \mathbf{P}_{\beta,c}^{(1)} \mathbf{1}_{5\times 1})^{-1} \mathbf{1}_{5\times 1}^{\mathrm{T}} \mathbf{P}_{\beta,c}^{(1)} \mathbf{1}_{\beta,c}; \qquad (20)$$

$$\widehat{\sigma_{1;\epsilon,c}^{2(1)}} \equiv m_{1;\epsilon,c}^{2(1)} = (\mathbf{I}_{5\times 1}^{\mathrm{T}} \mathbf{P}_{\epsilon,c}^{(1)} \mathbf{1}_{5\times 1})^{-1} \mathbf{I}_{5\times 1}^{\mathrm{T}} \mathbf{P}_{\epsilon,c}^{(1)} \mathbf{1}_{\epsilon,c} .$$
(21)

Step 5. Repeating steps 3 and 4 until the following conditions are met (τ denotes an iteration number):

$$\left| m_{\mathbf{l};\alpha(\beta)(\varepsilon),c}^{2\ (\tau+1)} - m_{\mathbf{l};\alpha(\beta)(\varepsilon),c}^{2\ (\tau)} \right| \leq 10^{-1};$$

$$(22)$$

$$m_{2;\alpha,c}^{2\ (\tau+1)} - m_{2;\alpha,c}^{2\ (\tau)} \bigg| \le 10^{-1}.$$
(23)

Step 6. Adopting final estimates $m_{1;\alpha,c}^2$, $m_{2;\alpha,c}^2$, $m_{1;\beta,c}^2$ and $m_{1;\epsilon,c}^2$, with the corresponding degrees of freedom, calculated as $f_{1,2;\alpha,c} = \sum_{s=1}^{5} f_{\alpha,c;s}$ (for the first two), $f_{1;\beta,c} = \sum_{s=1}^{5} f_{\beta,c;s}$ and $f_{1;\epsilon,c} = \sum_{s=1}^{5} f_{\epsilon,c;s}$. Degrees of freedom

 $f_{\xi,c;s}$, $\xi \in \{\alpha,\beta,\epsilon\}$, are calculated as shown in Anđić (2016, 2019a).

2. Results

In Tables 1 to 6, the results of ANOVA estimation performed for the daytime and nighttime within the fouryear period considered (2008–2011), are given. The results are square roots of epoch-wise variance components estimates (i.e. standard deviations estimates) obtained using individually monthly datasets including daytime- and those including nighttime-wise true errors. In doing so, a linear model based on two-way nested classification with random effects with no interactions was used (see Anđić, 2016, 2019a).

Square roots of the final epoch-wise variance components estimates, by months within the period 2008–2011, are presented graphically in Figures 2 to 5.

Baseline	$m^{DT}_{\varepsilon,e}$	[mm]	$m^{DT}_{\beta,e}[mm]$		$m^{DT}_{lpha,e}$ [mm]	
Dusenne	min	max	min	max	min	max
TORI-IENG	2.1	2.9	2.5	4.9	1.4	4.8
IGMI-PRAT	3.2	4.1	2.5	5.0	1.8	4.7
BAR-PODG	3.2	4.9	3.4	6.1	2.2	6.6
BOR1-WROC	2.6	3.3	2.7	5.7	1.9	7.4
BACA-BAIA	2.6	3.5	4.0	15.2	4.7	235.6

Table 1. Extreme values of epoch-wise standard deviations estimates for the coordinate *e* and *the daytime* (*DT*) *period*

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Table 2. Extreme values of epoch-wise standard deviations estimates for the coordinate n and the daytime (DT) period

Baseline	$m_{\varepsilon,n}^{DT}$	[mm]	$m^{DT}_{eta,n}$ [mm]		$m^{DT}_{\alpha,n \ [mm]}$		
Dusenne	min	max	min	max	min	max	
TORI-IENG	2.5 3.9 2.9		8.3	1.5	8.4		
IGMI-PRAT	3.7	5.7	2.9	7.1	1.9	6.6	
BAR-PODG	4.2	6.6 4.4 7.		7.9	2.3	9.2	
BOR1-WROC	3.5 4.8 3.6 4.8		3.1	7.5	3.9	10.8	
BACA-BAIA			5.6	16.8	4.3	131.4	

Table 3. Extreme values of epoch-wise standard deviations estimates for the coordinate *u* and *the daytime* (*DT*) *period*

Baseline	$m_{\varepsilon,u}^{DT}$	[mm]	$m_{\beta,u\ [mm]}^{DT}$ m_{α}^{I}			DT .,u [mm]	
Dusenne	min	max	min	max	min	max	
TORI-IENG	5.1	8.0	6.8	16.6	3.7	16.2	
IGMI-PRAT	8.8	11.0	7.4	14.4	6.1	23.0	
BAR-PODG	8.2	13.5 9.1 1		17.0	13.5	38.7	
BOR1-WROC	6.8	8.5	7.3 15.7		16.4	53.6	
BACA-BAIA	6.5 8.		10.5	40.6	33.1	326.4	

Table 4. Extreme values of epoch-wise standard deviations estimates for the coordinate e and the nighttime (NT) period

Baseline	$m_{\varepsilon,e}^{NT}$	[<i>mm</i>]	$m^{NT}_{\beta,e}$	$m^{NT}_{\beta,e\ [mm]}$ $m^{NT}_{\alpha,e\ [mm]}$		
	min	max	min	max	min	max
TORI-IENG	2.1	2.1 2.7 2.3 4.8		4.8	1.4	4.6
IGMI-PRAT	3.2 3.9		2.2	4.4	1.1	4.1
BAR-PODG	3.3	4.7	3.4	5.3	1.6	4.3
BOR1-WROC	2.6	3.1	2.7	4.4	1.8	5.9
BACA-BAIA	2.7 3.1		3.9	7.8	4.3	157.6

Table 5. Extreme values of epoch-wise standard deviations estimates for the coordinate *n* and *the nighttime* (*NT*) *period*

Baseline	$m_{\varepsilon,n}^{NT}$	[<i>mm</i>]	$m^{NT}_{\beta,n \ [mm]}$		$m^{NT}_{\alpha,n \ [mm]}$		
	min	max	min	max	min	max	
TORI-IENG	2.4	.4 3.3 3.1 8.0		8.0	1.7	6.9	
IGMI-PRAT	4.0 5.6		3.1	5.8	2.1	5.1	
BAR-PODG	4.4	6.1	4.5	6.5	2.2	5.5	
BOR1-WROC	3.6	4.4	3.8	5.6	3.7	14.0	
BACA-BAIA	3.7	4.3	6.3	12.9	5.0	125.2	

Table 6. Extreme values of epoch-wise standard deviations estimates for the coordinate *u* and *the nighttime* (*NT*) *period*

Baseline	$m_{\varepsilon,u}^{NT}$	[mm]	$m_{\beta,u}^{NT}$	[mm]	$m^{NT}_{\alpha,u \ [mm]}$		
	min	max	min	max	min	max	
TORI-IENG	5.3 6.4		6.4	15.7	2.9	16.3	
IGMI-PRAT	8.7 11.3		7.3	12.8	7.2	16.7	
BAR-PODG	8.0	13.1 9.0		14.8	12.7	35.3	
BOR1-WROC	6.5	8.0	6.7	12.8	15.1	62.4	
BACA-BAIA	6.5	6.5 8.0		27.0	24.0	237.9	



Figure 2. Final epoch-wise standard deviation estimates by the relative coordinates (*e*, *n* and *u*) for the purely random error, obtained *for the daytime and the nighttime period*



Figure 3. Final epoch-wise standard deviation estimates by the relative coordinates (*e*, *n* and *u*) for the effect of FF MP, obtained *for the daytime and the nighttime period*



Figure 4. Final epoch-wise standard deviation estimates by the relative coordinates (*e*, *n* and *u*) for the joint effect of NF MP and RAPCOV, obtained *for the daytime and the nighttime period*



Figure 5. Final epoch-wise standard deviation estimates by the relative coordinates (*e*, *n* and *u*) for the joint effect of TI, obtained *for the daytime and the nighttime period*

After applying the well known *F-test for equality of two variances* (for details, a reader can see, e.g. Snedecor & Cochran, 1989), herein based on the testing:

(a)
$$H_{0;1}: \sigma_{1;\alpha(\beta),c;DT}^2 = \sigma_{1;\alpha(\beta),c;NT}^2$$
, (24)

against $H_{a1;1}:\sigma_{1;\alpha(\beta),c;DT}^2 > \sigma_{1;\alpha(\beta),c;NT}^2$,

for
$$m_{1;\alpha(\beta),c;DT}^2 > m_{1;\alpha(\beta),c;NT}^2$$
, (25)

as well as:

$$\begin{split} H_{a2;1} : &\sigma_{1;\alpha(\beta),c;DT}^2 < \sigma_{1;\alpha(\beta),c;NT}^2 ,\\ &\text{for } m_{1;\alpha(\beta),c;DT}^2 < m_{1;\alpha(\beta),c;NT}^2 , \end{split}$$
(26)

and also:

(b)
$$H_{0;2}:\sigma_{2;\alpha,c;DT}^2 = \sigma_{2;\alpha,c;NT}^2$$
, against (27)

$$H_{a1;2}:\sigma_{2;\alpha,c;DT}^2 > \sigma_{2;\alpha,c;NT}^2, \text{ for } m_{2;\alpha,c;DT}^2 > m_{2;\alpha,c;NT}^2, (28)$$

as well as:

$$H_{a2;2}:\sigma_{2;\alpha,c;DT}^2 < \sigma_{2;\alpha,c;NT}^2, \text{ for } m_{2;\alpha,c;DT}^2 < m_{2;\alpha,c;NT}^2, (29)$$

(*DT* and *NT* were introduced to denote daytime and nighttime).

So, adopting the 5% significance level, one obtained the results that are given in Table 7.

Conclusions

This study, based on a very large and representative sample, showed that there is statistically significant inequality when comparing corresponding variances of unmodelled effects (limiting relative GPS positioning accuracy) obtained for daytime and nighttime periods (see Table 7).

When it comes to extreme values of the final epochwise standard deviation estimates for the effects considered, one can be spotted that a seasonal pattern is also present, especially pronounced with the joint effect of TI (see Figures 2 to 5). These extreme values (with the related degrees of freedom), as well as the corresponding mean values of those standard deviations estimates are given in the continuation by the relative coordinates (*e*, *n* and *u*).

"Far-field" multipath effect:

Daytime period

$$-\min m_{1:\beta,e}^{DT} = 3.3 \,\mathrm{mm} \,(19198.65 \,\mathrm{d.f});$$

 $\max m_{1;\beta,e}^{DT} = 6.9 \text{ mm} (21027.77 \text{ d.f});$ $mean m_{1;\beta,e}^{DT} = 4.4 \text{ mm};$

- min $m_{l;\beta,n}^{DT}$ = 4.6 mm (12928.44 d.f); max $m_{l;\beta,n}^{DT}$ = 9.0 mm (20815.24 d.f); mean $m_{l;\beta,n}^{DT}$ = 6.0 mm;

$$-\min m_{1;\beta,u}^{DT} = 9.1 \text{ mm (19106.55 d.f)};$$

$$\max m_{1;\beta,u}^{DT} = 20.3 \text{ mm (22763.77 d.f)};$$

$$\operatorname{mean} m_{1;\beta,u}^{DT} = 12.5 \text{ mm.}$$

Nighttime period

$$-\min m_{1;\beta,e}^{NT} = 3.2 \text{ mm (13406.56 d.f)};$$
$$\max m_{1;\beta,e}^{NT} = 4.9 \text{ mm (16609.49 d.f)};$$
$$\operatorname{mean} m_{1;\beta,e}^{NT} = 4.0 \text{ mm};$$

- $\min m_{1;\beta,n}^{NT} = 4.7 \text{ mm} (18597.80 \text{ d.f});$ $\max m_{1;\beta,n}^{NT} = 7.3 \text{ mm} (18783.11 \text{ d.f});$ $\operatorname{mean} m_{1;\beta,n}^{NT} = 5.6 \text{ mm};$
- $-\min m_{1;\beta,u}^{NT} = 8.4 \text{ mm (16092.83 d.f)};$ $\max m_{1;\beta,u}^{NT} = 15.4 \text{ mm (19477.51 d.f)};$ $\operatorname{mean} m_{1;\beta,u}^{NT} = 11.3 \text{ mm.}$

Joint effect of "near-field" multipath and receiver antenna phase center offset and variations:

Daytime period

- min $m_{1;\alpha,e}^{DT}$ = 1.5 mm (507.88 d.f); max $m_{1;\alpha,e}^{DT}$ = 4.7 mm (728.60 d.f); mean $m_{1;\alpha,e}^{DT}$ = 2.9 mm;
- min $m_{l;\alpha,n}^{DT}$ =1.9 mm (701.42 d.f); max $m_{l;\alpha,n}^{DT}$ =7.0 mm (856.69 d.f); mean $m_{l;\alpha,n}^{DT}$ = 4.5 mm;
- min $m_{1;\alpha,u}^{DT}$ = 3.4 mm (739.53 d.f); max $m_{1;\alpha,u}^{DT}$ = 21.9 mm (759.82 d.f); mean $m_{1;\alpha,u}^{DT}$ = 13.0 mm.

Nighttime period

 $-\min m_{l;\alpha,e}^{NT} = 0.8 \text{ mm } (691.85 \text{ d.f});$ $\max m_{l;\alpha,e}^{NT} = 3.8 \text{ mm } (802.20 \text{ d.f});$ $\operatorname{mean} m_{l;\alpha,e}^{NT} = 2.5 \text{ mm};$

	tage by the relative coordinates (e , n and P and RAPCOV (α 1) and the joint effect	
ß	<i>α</i> 1	a2

	β				α1		α2		
	е	п	и	е	п	и	е	п	и
H _{a1;1(2)}	81.3%	56.3%	83.3%	64.6%	62.5%	72.9%	46.7%	46.8%	39.6%
H _{0;1(2)}	4.2%	4.2%	8.3%	22.9%	14.6%	16.7%	11.1%	19.1%	12.5%
H _{a2;1(2)}	14.6%	39.6%	8.3%	12.5%	22.9%	10.4%	42.2%	34.0%	47.9%

- min $m_{l;\alpha,n}^{NT}$ = 2.1 mm (522.22 d.f); max $m_{l;\alpha,n}^{NT}$ = 5.0 mm (732.56 d.f); mean $m_{l;\alpha,n}^{NT}$ = 3.7 mm;
- min $m_{l;\alpha,u}^{NT}$ = 3.1 mm (687.14 d.f); max $m_{l;\alpha,u}^{NT}$ =15.8 mm (710.69 d.f); mean $m_{l;\alpha,u}^{NT}$ = 10.7 mm.

Joint effect of tropospheric and ionospheric refraction:

Daytime period

- $-\min m_{2;\alpha,e}^{DT} = 0.0116 \text{ mm/km} (798.60 \text{ d.f});$ $\max m_{2;\alpha,e}^{DT} = 0.3282 \text{ mm/km} (859.90 \text{ d.f});$ $\operatorname{mean} m_{2;\alpha,e}^{DT} = 0.1170 \text{ mm/km};$
- min $m_{2;\alpha,n}^{DT}$ =0.0103 mm/km (866.85 d.f); max $m_{2;\alpha,n}^{DT}$ =0.2365 mm/km (856.69 d.f); mean $m_{2;\alpha,n}^{DT}$ = 0.1020 mm/km;
- $-\min m_{2;\alpha,u}^{DT} = 0.1222 \text{ mm / km (905.91 d.f)};$ $\max m_{2;\alpha,u}^{DT} = 0.7818 \text{ mm/km (944.97 d.f)};$ $\max m_{2;\alpha,u}^{DT} = 0.3675 \text{ mm/km}.$

Nighttime period

- min $m_{2;\alpha,e}^{NT}$ = 0.0118 mm/km (796.24 d.f); max $m_{2;\alpha,e}^{NT}$ = 0.2734 mm/km (538.58 d.f); mean $m_{2;\alpha,e}^{NT}$ = 0.1149 mm/km;
- min $m_{2;\alpha,n}^{NT}$ =0.0097 mm/km (792.19 d.f); max $m_{2;\alpha,n}^{NT}$ =0.2289 mm/km (846.56 d.f); mean $m_{2;\alpha,n}^{NT}$ = 0.0873 mm/km;
- min $m_{2;\alpha,u}^{NT}$ = 0.0752 mm/km (633.06 d.f); max $m_{2;\alpha,u}^{NT}$ = 0.6315 mm/km (853.08 d.f); mean $m_{2;\alpha,u}^{NT}$ = 0.3806 mm/km.

The results obtained in this study can be very useful in optimal planning of experiments, as well as in engineering works where relative GPS positioning is used.

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